

# Introduction to Linear Algebra

## Fundamental Vector Operations

### Lecture 10

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## why do we need linear algebra?

- **Coordinates and Geometry:**  
Linear algebra provides a way to represent and understand geometric concepts like lines, planes, and transformations in higher dimensions.
- **Equations Systems:**  
It offers tools to solve systems of linear equations, which appear in countless areas of study.

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## what is linear algebra?

### Linear

- having to do with lines, planes, space, etc.
- example:  $x + y + 3z = 7$  (and not  $\sin, \log, x^2, \dots$ )

### Algebra

- solving equations involving numbers and symbols
- "*reunion of broken parts*"

Linear algebra is the math of vectors and matrices, so we'll start by definitions and the mathematical operations we can perform on vectors and matrices.

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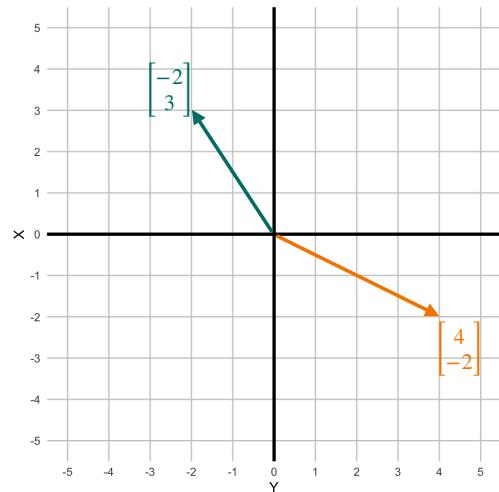
## what is a vector?

- object that has both a magnitude and a direction (physics)
- a list of numbers (data science)
- vector is rooted at **origin (0,0)**
- **components:** the entries of a vector
- the **dimension** of a vector is the number of components in the vector.
- notation: bold face (**x, y, z**) or arrow over ( $\vec{x}, \vec{y}, \vec{z}$ )



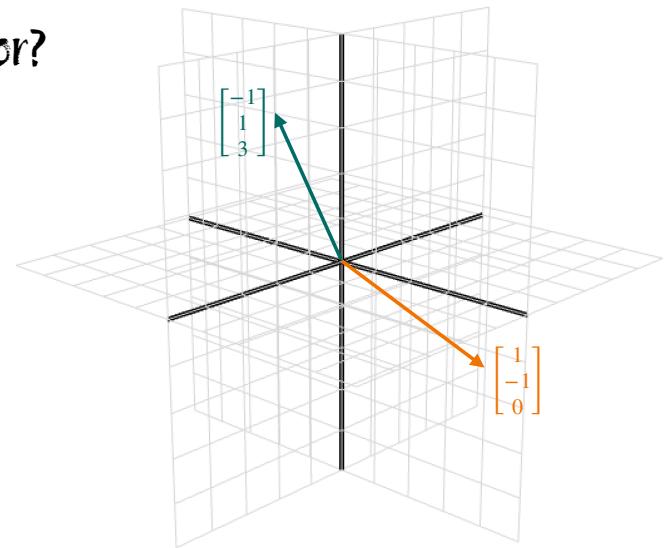
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## what is a vector?



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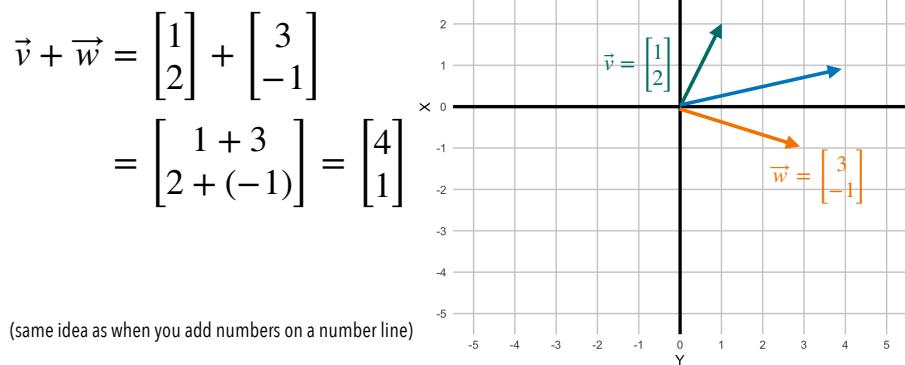
## what is a vector?



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## vector operations

### vector addition



(same idea as when you add numbers on a number line)

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## vector operations

### scalar multiplication

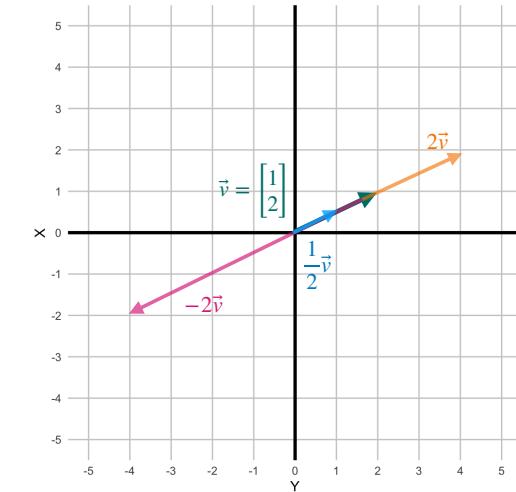
the process of stretching and squishing is called **scaling**

the **scalars** here are 2, -2 and  $\frac{1}{2}$

$$2\vec{v} = 2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

$$-2\vec{v} = -2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} -2 \\ -4 \end{bmatrix}$$

$$\frac{1}{2}\vec{v} = \frac{1}{2} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 0.5 \\ 1 \end{bmatrix}$$



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## magnitudes and direction

### example

What is the magnitude of  $\vec{a} = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$ ?

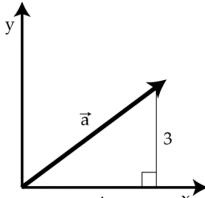
The **magnitude** of a vector is the distance from the endpoint of the vector to the origin, i.e. its **length**

This vector extends 4 units along the x-axis, and 3 units along the y-axis.

Magnitude  $\|\vec{a}\|$  is computed using Pythagorean Theorem ( $x^2 + y^2 = z^2$ ):

$$\|\vec{a}\| = \sqrt{x^2 + y^2} = \sqrt{4^2 + 3^2} = 5$$

The magnitude of a vector is a scalar value.



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## basis vectors

the  $x$  coordinate of a vector can be viewed as a scalar scaling  $\hat{i}$

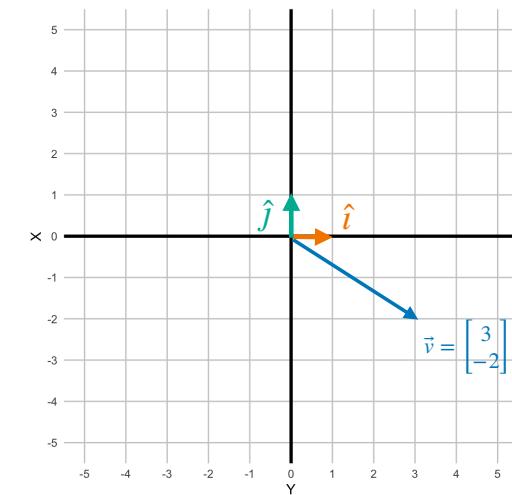
the  $y$  coordinate of a vector can be viewed as a scalar scaling  $\hat{j}$

$\hat{i}$  and  $\hat{j}$  are **unit vectors**: they have length 1

$$\vec{v} = (3)\hat{i} + (-2)\hat{j}$$

$(\hat{i}, \hat{j})$  are called the **basis vectors** of the  $xy$  coordinate system

Any vector in the plane can be written as a combination of these two vectors



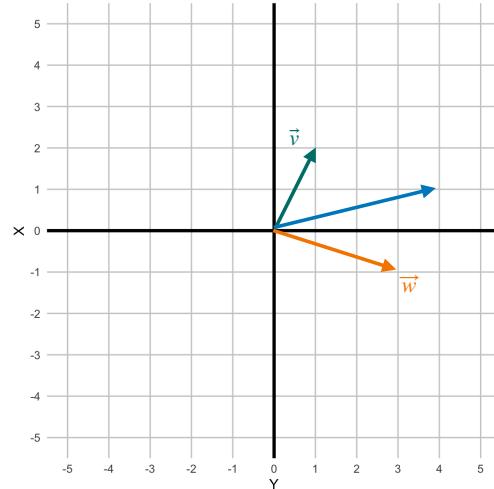
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## basis vectors

what if we chose some other basis vector?

the choice of basis vector effect the numerical values of the vectors

scaling vectors and adding them is a **linear combination** of those two vectors



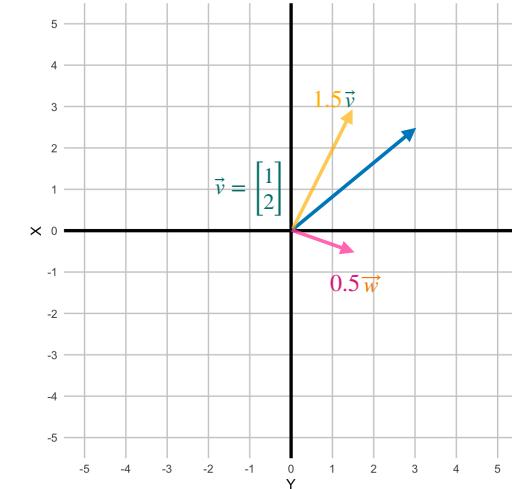
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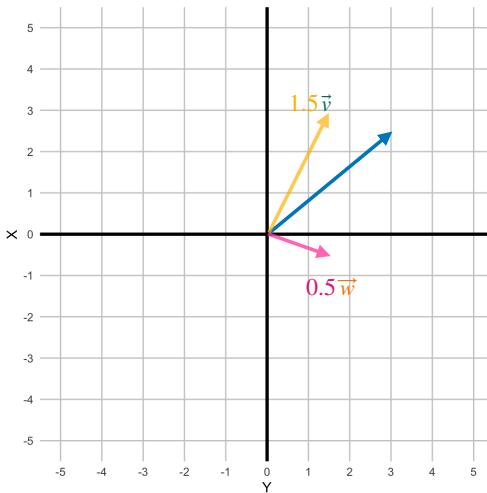
the choice of basis vector effect the numerical values of the vectors

scaling vectors and adding them is called a **linear combination** of those two vectors

**the span** of the vector  $\vec{v}$  and  $\vec{w}$  is the set of all their linear combinations

$$a\vec{v} + b\vec{w}$$

where  $a, b \in \mathbb{R}$ .



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## basis vectors

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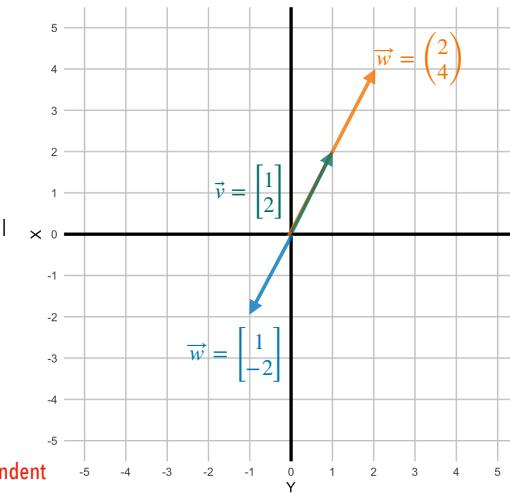
**the span** of the vector  $\vec{v}$  and  $\vec{w}$  is the set of all their linear combinations

$$a\vec{v} + b\vec{w}$$

where  $a, b \in \mathbb{R}$ .

If  $\vec{v}$  and  $\vec{w}$  do not line up (they are not scalar multiples of each other), their span is a plane. If they do line up, their span is a line.

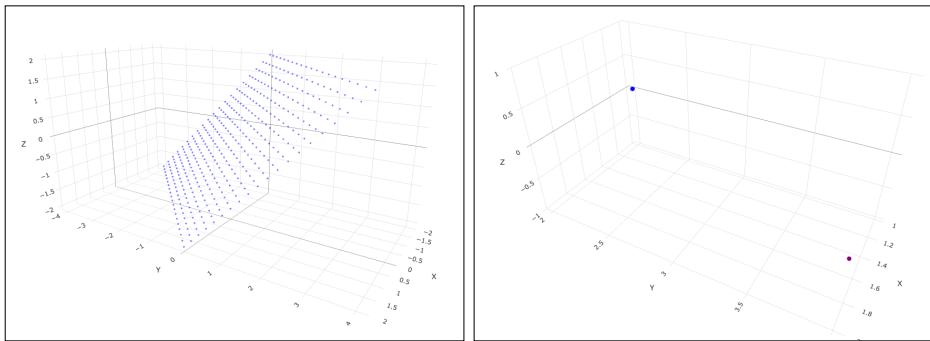
In the latter case the vectors are **linearly dependent**



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## linear combinations: planes and lines

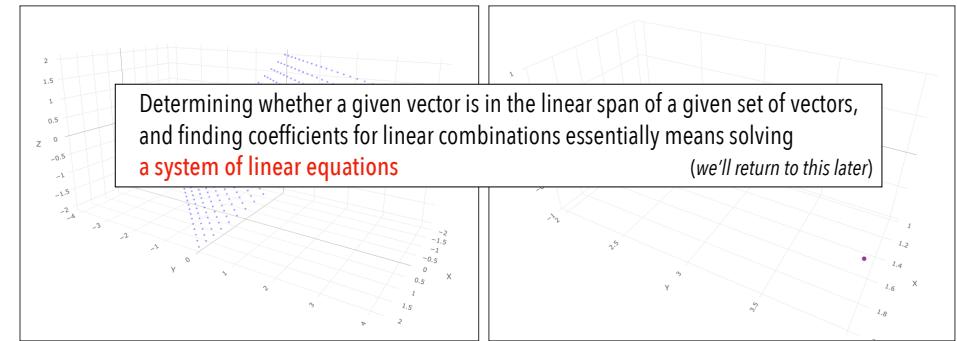
- Scaling vectors and adding them is a **linear combination** of those two vectors
- Geometrically**, the linear combinations of a nonzero vector form a line. The linear combinations of two nonzero vectors form a plane, unless the two vectors are collinear, in which case they form a line.



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## linear combinations: planes and lines

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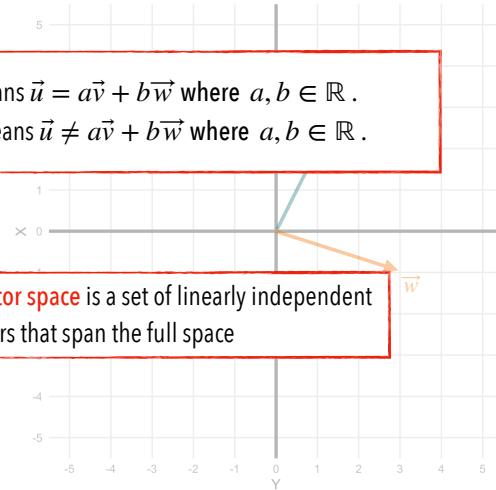


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## linear independence

Linearly dependent means  $\vec{u} = a\vec{v} + b\vec{w}$  where  $a, b \in \mathbb{R}$ .

Linearly independent means  $\vec{u} \neq a\vec{v} + b\vec{w}$  where  $a, b \in \mathbb{R}$ .



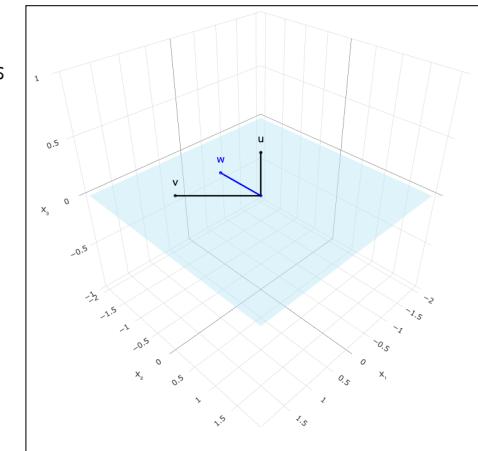
the **basis of the vector space** is a set of linearly independent vectors that span the full space

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## linear independence and spanning vectors

$\vec{w}$  is in  $\text{span}(\vec{u}, \vec{v})$  or the plane spanned by  $(\vec{u}, \vec{v})$

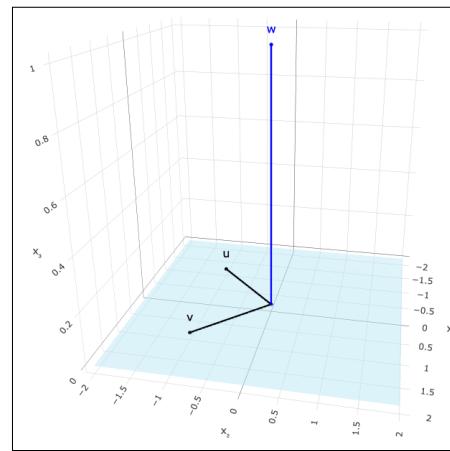
$\vec{w}$  is a linear combination of  $(\vec{u}, \vec{v})$ , so  $(\vec{u}, \vec{v}, \vec{w})$  is not linear independent.



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## linear independence and spanning vectors

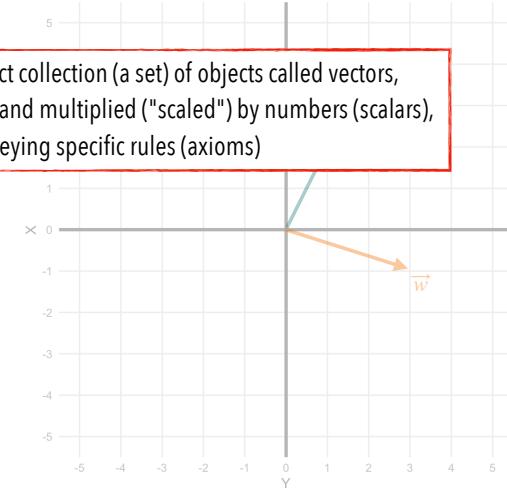
Since  $\vec{w}$  is not in  $\text{span}(\vec{u}, \vec{v})$ ,  $(\vec{u}, \vec{v}, \vec{w})$  is linear independent.



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## the vector space

A **vector space** is an abstract collection (a set) of objects called vectors, which can be added together and multiplied ("scaled") by numbers (scalars), all while obeying specific rules (axioms)



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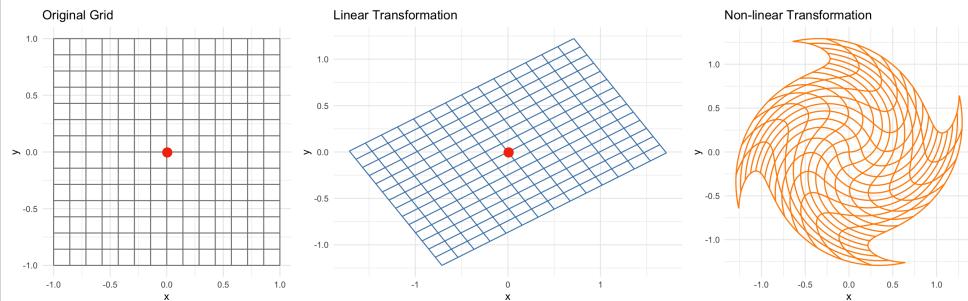
## the vector space

The 'rules' (axioms) to hold true for all vectors  $\vec{u}, \vec{v}, \vec{w} \in V$  and scalars  $a, b$  are the following

- Closure:  $\vec{u} + \vec{v} \in V, a\vec{v} \in V$
- Commutative:  $\vec{u} + \vec{v} = \vec{v} + \vec{u}$
- Associativity:  $(\vec{u} + \vec{v}) + \vec{w} = \vec{u} + (\vec{v} + \vec{w})$
- Identity for Addition: There is a zero vector  $\vec{0} \in V$  such that  $\vec{u} + \vec{0} = \vec{u}$
- Distributive Property:  $a(\vec{u} + \vec{v}) = a\vec{u} + a\vec{v}$  and  $(a + b)\vec{u} = a\vec{u} + b\vec{u}$
- Associativity of Scalars:  $(ab)\vec{u} = a(b\vec{u})$
- Multiplicative Identity:  $1\vec{u} = \vec{u}$

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## linear transformations & matrices



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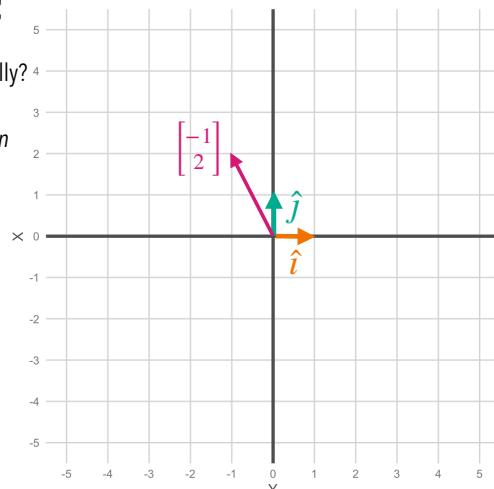
## linear transformations

how to describe linear transformations numerically?

by using unit vectors  $\hat{i}$  and  $\hat{j}$  and keeping track of where they land after the transformation

example

$$\vec{v} = -1\hat{i} + 2\hat{j}$$



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## linear transformations

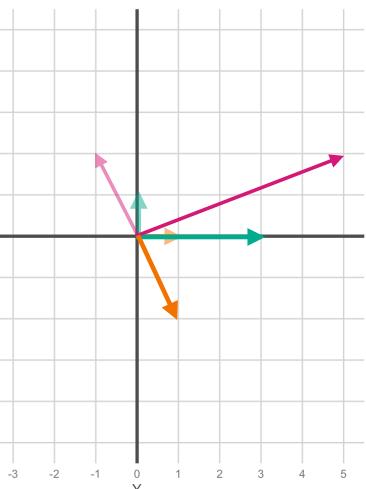
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example

$$\vec{v} = -1\hat{i} + 2\hat{j}$$

$$\begin{aligned}\vec{v}_{\text{transformed}} &= -1\hat{i}_{\text{transformed}} + 2\hat{j}_{\text{transformed}} \\ &= -1 \begin{bmatrix} 1 \\ -2 \end{bmatrix} + 2 \begin{bmatrix} 3 \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} -1(1) + 2(3) \\ -1(-2) + 2(0) \end{bmatrix} = \begin{bmatrix} 5 \\ 2 \end{bmatrix}\end{aligned}$$



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## linear transformations

how to describe linear transformations numerically?

by using unit vectors  $\hat{i}$  and  $\hat{j}$  and keeping track of where they land after the transformation

generally

$$\hat{i} \rightarrow \begin{bmatrix} 1 \\ -2 \end{bmatrix} \quad \hat{j} \rightarrow \begin{bmatrix} 3 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix} \rightarrow x \begin{bmatrix} 1 \\ -2 \end{bmatrix} + y \begin{bmatrix} 3 \\ 0 \end{bmatrix} = \begin{bmatrix} 1x + 3y \\ -2x + 0y \end{bmatrix}$$



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any 2D linear transformation is completely described using only these four numbers

let's put them in a  $2 \times 2$  grid aka a matrix

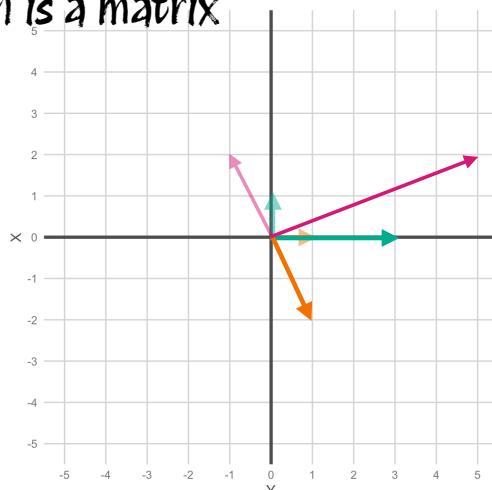


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## a linear transformation is a matrix

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

where  $\hat{i}$  lands      where  $\hat{j}$  lands



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## matrix as a transformation of space

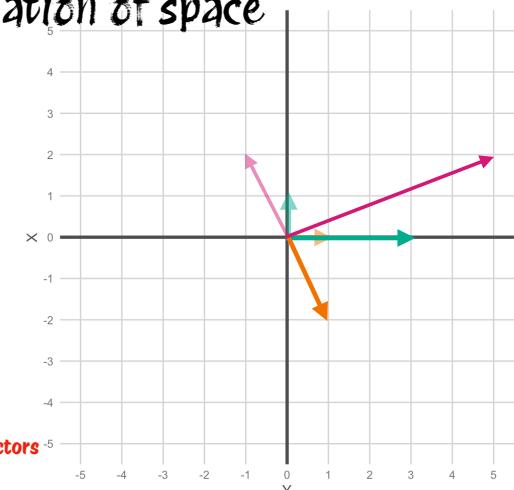
$$\begin{bmatrix} 1 & 3 \\ -2 & 0 \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix}$$

any vector

$$\Rightarrow x \begin{bmatrix} 1 \\ -2 \end{bmatrix} + y \begin{bmatrix} 3 \\ 0 \end{bmatrix}$$

adding scaled versions of our new basis vectors



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## matrix as a transformation of space

generally:

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\begin{aligned} & \begin{bmatrix} x \\ y \end{bmatrix} \\ \implies x \begin{bmatrix} a \\ c \end{bmatrix} + y \begin{bmatrix} b \\ d \end{bmatrix} &= \begin{bmatrix} ax + by \\ cx + dy \end{bmatrix} = \underbrace{\begin{bmatrix} a & b \\ c & d \end{bmatrix}}_{\text{matrix vector multiplication}} \begin{bmatrix} x \\ y \end{bmatrix} \end{aligned}$$

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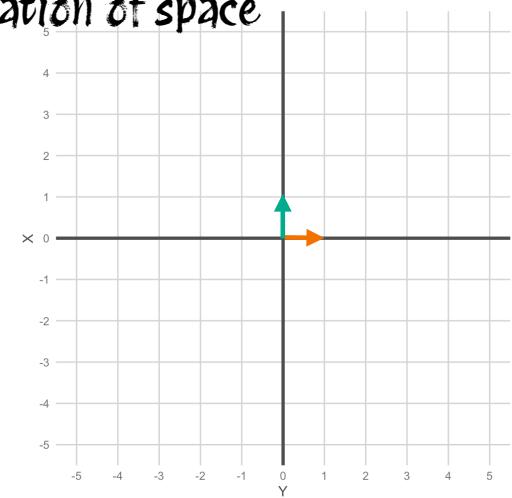
## matrix as a transformation of space

example

rotate space 90° counterclockwise

$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

where  $i$  lands      where  $j$  lands



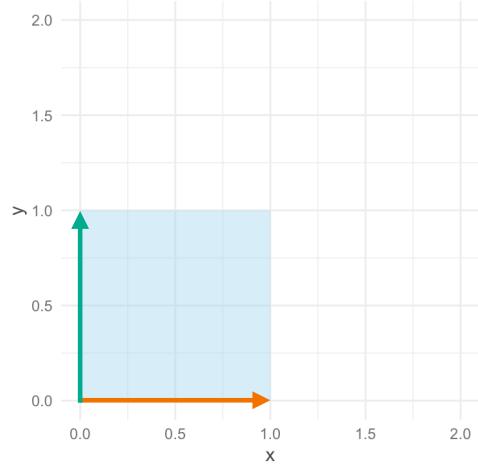
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## matrix as a transformation of space

example

the **identity matrix** (no transformation)

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
$$= \begin{bmatrix} 1x + 0y \\ 0x + 1y \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$



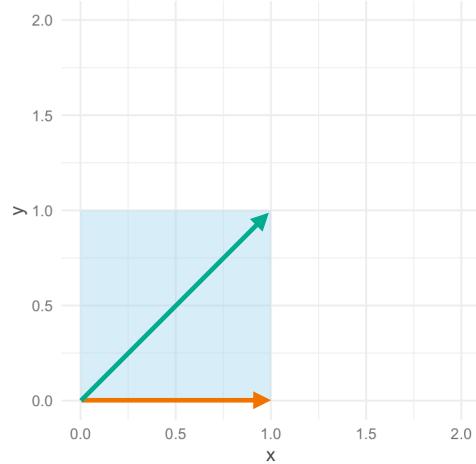
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## matrix as a transformation of space

example

a **shear** transformation

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



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## matrix as a transformation of space

### example

a **shear** transformation

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1x + 1y \\ 0x + 1y \end{bmatrix} = \begin{bmatrix} x + y \\ y \end{bmatrix}$$

