

VISPAD INSTITUTE OF  
TECHNOLOGY  
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# Vectors

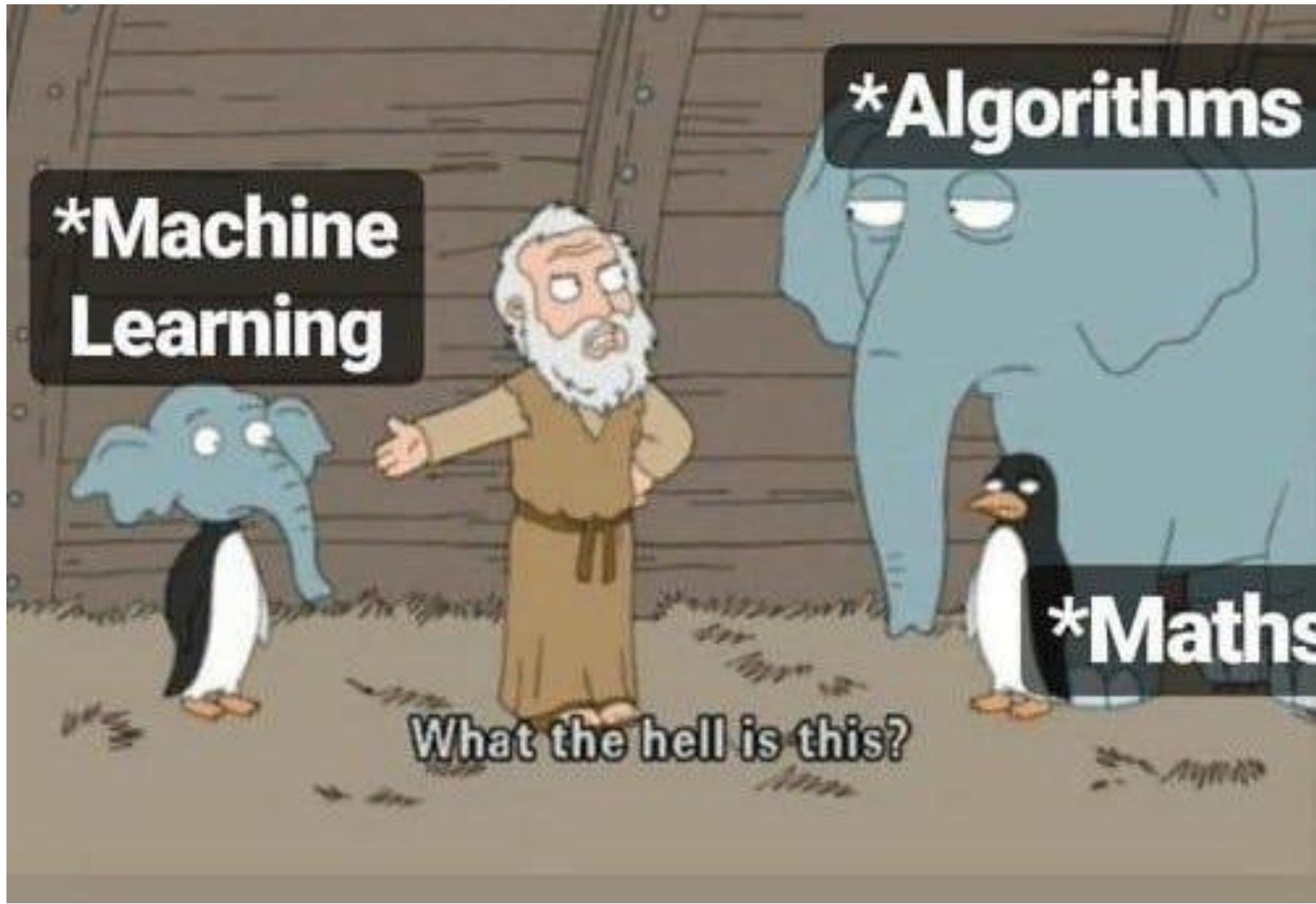


Fall 2025   
(October – December Virtual Internship)

**At the end of the session the students or candidates should be able to understand and work with:**



- Vector spaces
- Basis
  - Linear combination and independence
  - Spans of vectors
  - Bases
  - Finite dimensional vectors spaces
  - Why bases so important
  - The existences of bases
  - Sub spaces



\*Algorithms

\*Machine  
Learning

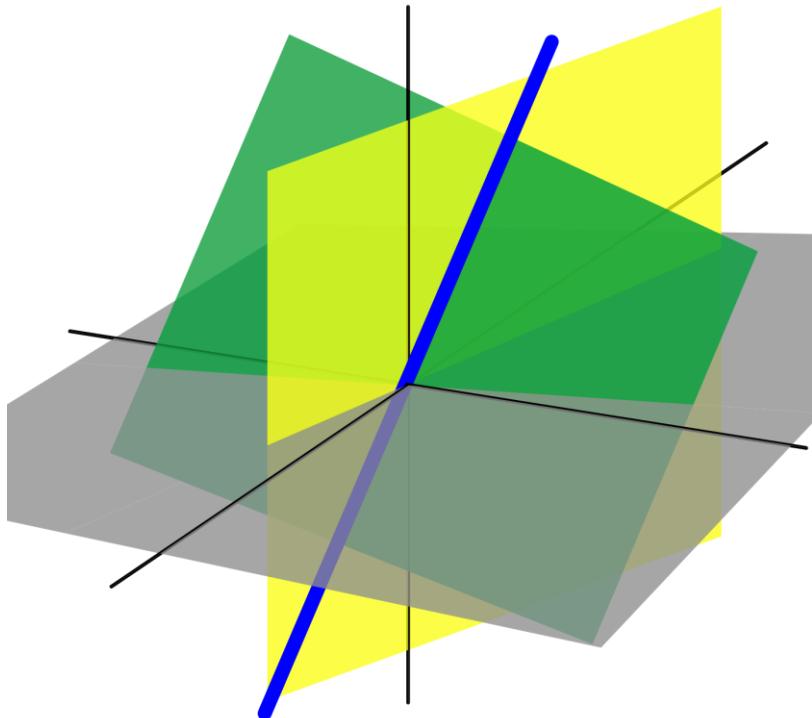
\*Maths

What the hell is this?



- Mathematics of machine learning rests upon three pillars:
  - linear algebra
  - calculus
  - probability theory
- Linear algebra describes how to **represent and manipulate data**; calculus helps us **fit the models**; while **probability theory helps interpret them**.

# Why Linear Algebra



- Linear algebra is essential to forming a **complete** understanding of machine learning.
- The techniques from this discipline belong to a shared collection of algorithm widely used in **artificial intelligence**.

- Its properties and methods allow for faster computation of complex systems and extraction of hidden relationships in sets of data.

# Notation

- $\mathbb{Z}, \mathbb{N}, \mathbb{Q}, \mathbb{R}, \mathbb{C}$  denote, respectively, the **integers**, the **natural numbers (nonnegative integers)**, the **rational numbers**, the **real numbers**, and the **complex numbers**.
- $\sum_{i=1}^n a_i = a_1 + a_2 + \dots + a_n$ ,  $\prod_{i=1}^n a_i = a_1 a_2 \dots a_n$ , are used for the sum and product of the quantities  $a_1, \dots, a_n$

# What is a Vector

- In a simple term, *a vector is a list of numbers where the position of each item in this structure matters.*
- A *column vector* consists of a finite number of real numbers, known as its entries, arranged in a vertical column.

□ Given a position integer  $n = 1, 2, 3, \dots$ , the set of all vectors with  $n$  entries is denoted by  $\mathbb{R}^n$ .

□ Examples of vectors in  $\mathbb{R}^3$ :

$$\begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix}, \begin{pmatrix} \pi \\ \sqrt{2} \\ -\frac{4}{7} \end{pmatrix}, \begin{pmatrix} 3.14 \\ 1.41 \\ -.57 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

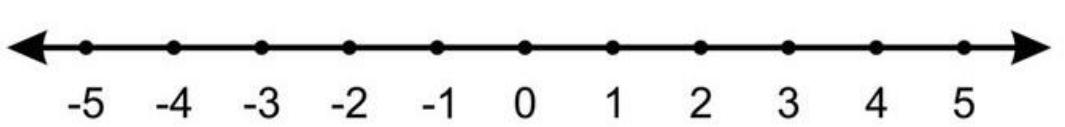
□ In general, a vector  $v \in \mathbb{R}^n$  has the form:  $v =$

$$\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix},$$

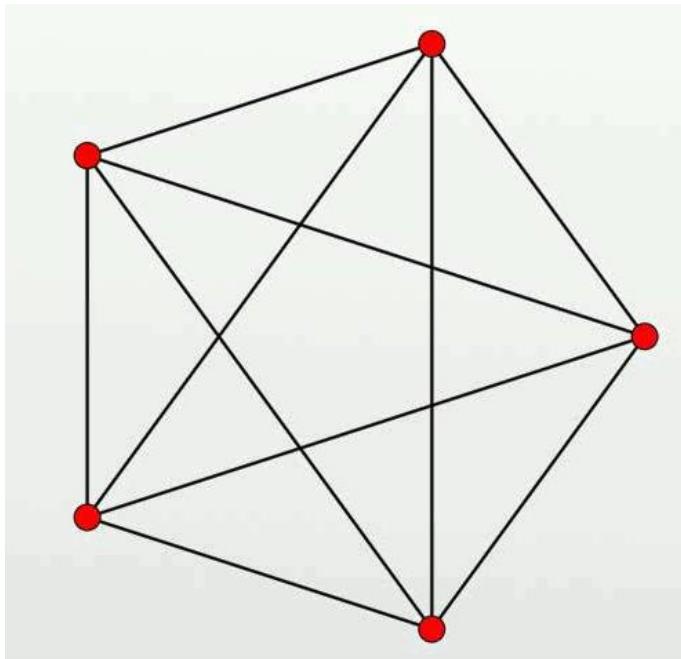
where  $v_1, \dots, v_n \in \mathbb{R}$ .

□ Two vectors are equal,  $v = w$ , if and only if they have the same of entries, so  $v, w \in \mathbb{R}^n$  for some  $0 < n \in \mathbb{N}$ , and all their entries are equal:  $v_i = w_i, i = 1, \dots, n$ .

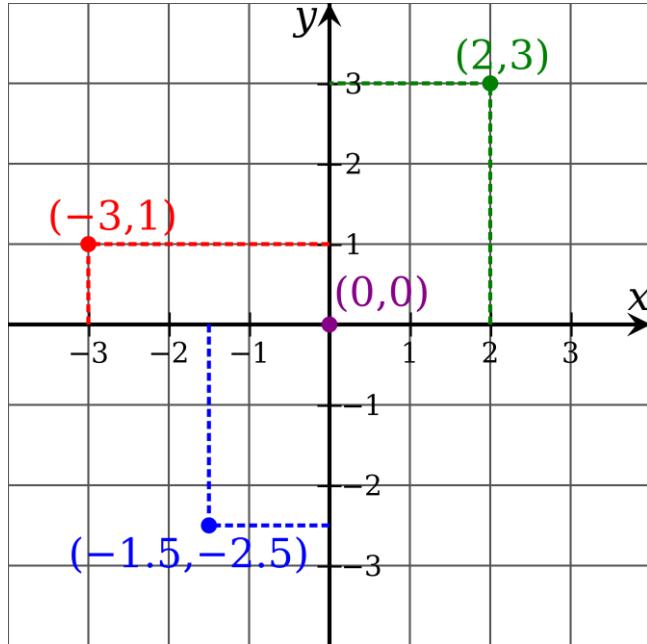
- The set  $\mathbb{R}^n$  is known as  $n -$ dimensional Euclidean space, which forms the basic setting for Euclidean geometry
- $\mathbb{R}^1 \simeq \mathbb{R}$  can be identified as the real line;
- $\mathbb{R}^2$  is the two-dimensional Euclidean plane;
- $\mathbb{R}^3$  can be identified with three-dimensional space



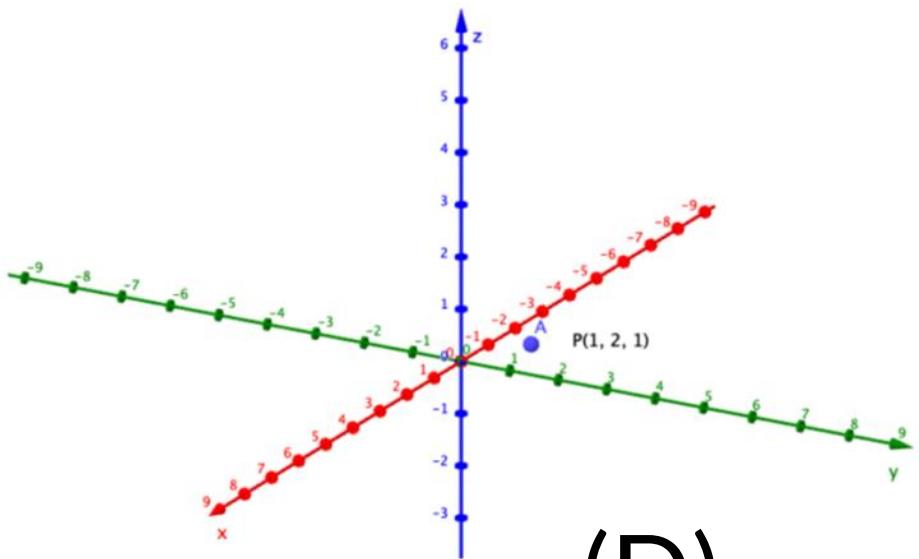
(A)



(B)



(C)



(D)

- A *row vector* contains a finite number of real numbers arranged in a horizontal row.
- *Column vectors* are the more important of the two.
- The operation of converting a column vector into a row vector, and vice versa, is known as the *transpose*.
- It is denoted with a  $T$  superscript.

$$\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}^T = (v_1, v_2, \dots, v_n)$$

$$(v_1, v_2, \dots, v_n)^T = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}.$$

Transposing twice takes you back to where you started:

$$(\boldsymbol{v}^T)^T$$

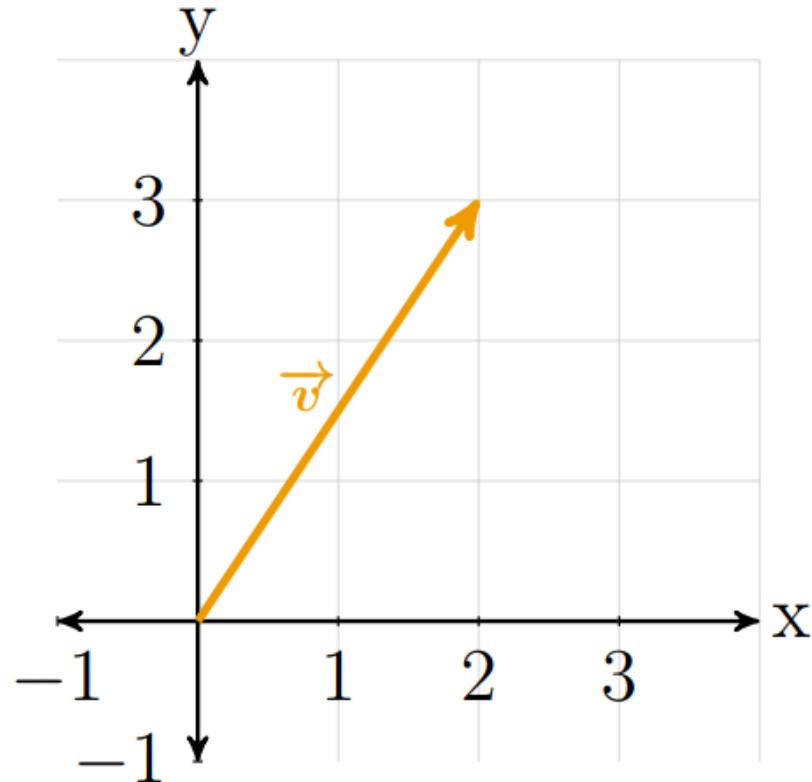
# Example

If you are analysing the **height** and **weight** of a class of students, in this domain, a **two-dimensional vector** will represent each student.

$$\vec{v} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 1.64 \\ 64 \end{pmatrix}$$

$v_1$  represent height,  $v_2$  represents weight;

□ In linear algebra, a vector is an arrow with a direction dictated by its coordinates.

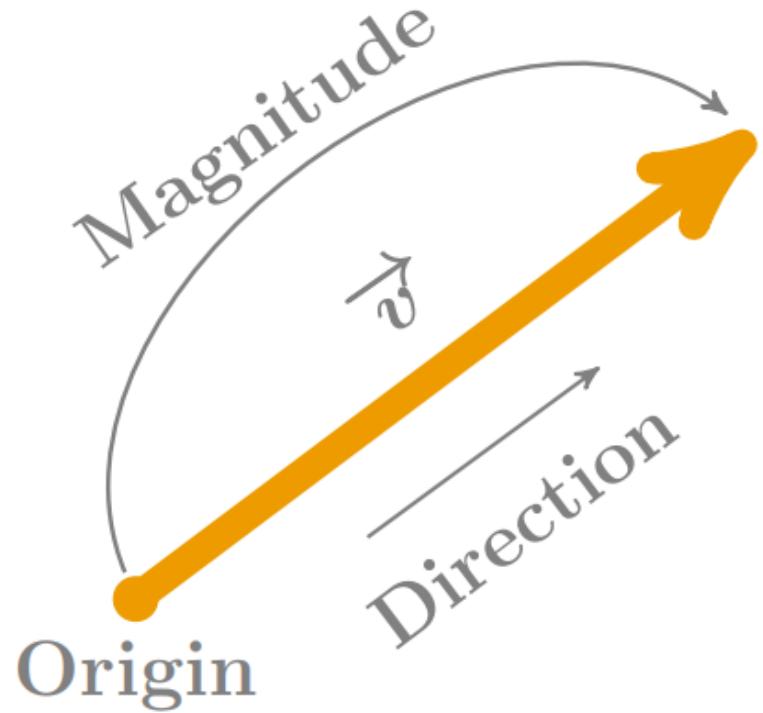


$$\vec{v} = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$

- There is no limitation to the dimensions a vector can have:
- A three-dimensional vector,  $t$  can be represented as:

$$\vec{t} = \begin{pmatrix} 2 \\ 3 \\ 5 \end{pmatrix}$$

- Using abstract concept, a vector is an object that has both a direction and a magnitude.
- The starting point of a vector is called the **BASE** or **ORIGIN**.



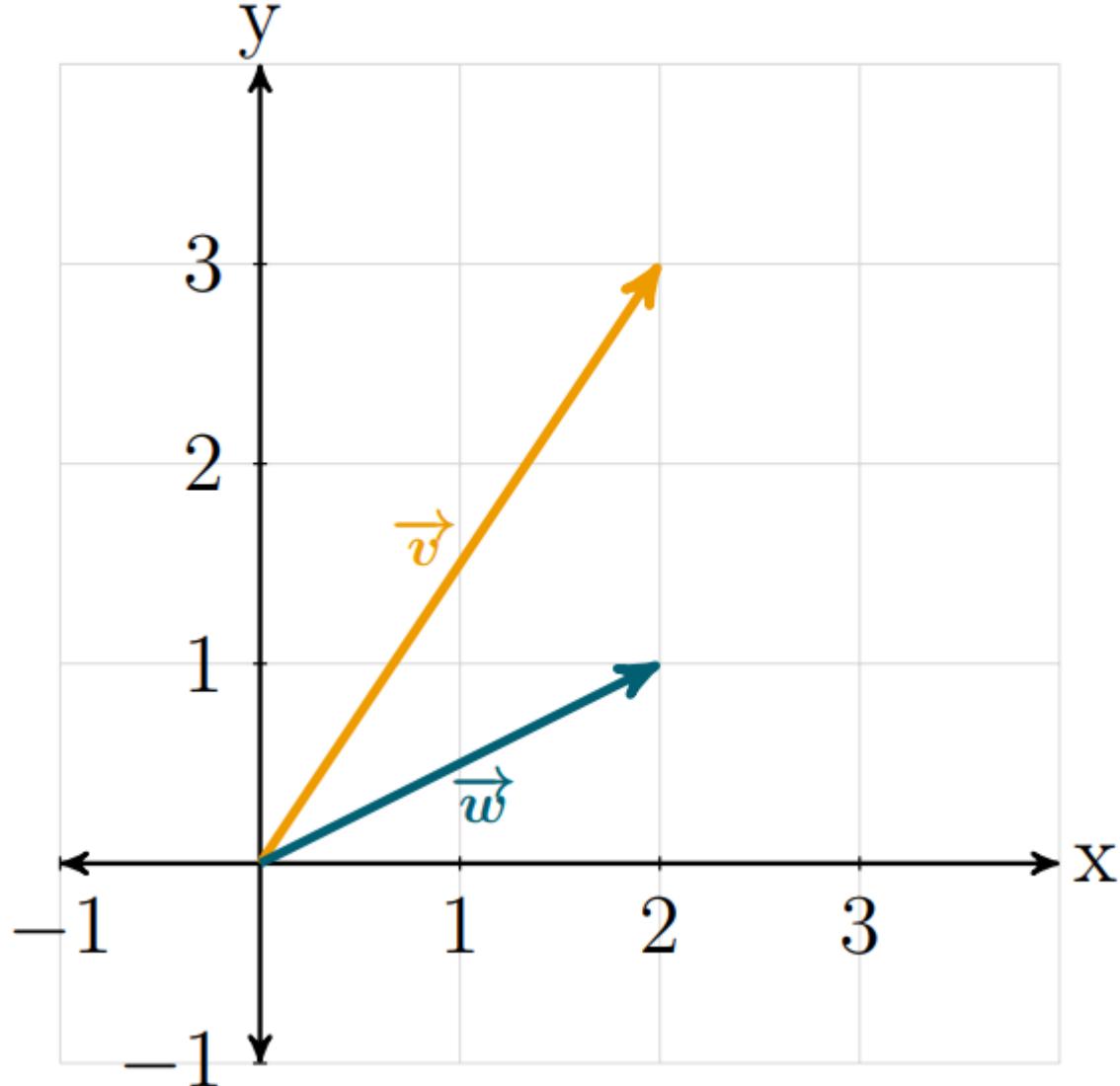
- ❑ **Magnitude** is the size of the vector, and it is also a function of where it lands.
- ❑ For two vectors to be the same, their **directions** and **magnitudes** must be equal.

- ❑ Vectors alone are instrumental mathematical elements.
- ❑ They can also represent many things, such as gravity, velocity, acceleration, and paths.

- When  $n = 1$ , a column vector  $\boldsymbol{v} = (v_1) \in \mathbb{R}^1$  has a single entry.
- Such a vector can be uniquely identified with the corresponding real number  $v_1 \in \mathbb{R}$ , and so  $\mathbb{R}^1 \simeq \mathbb{R}$

# Vector Addition

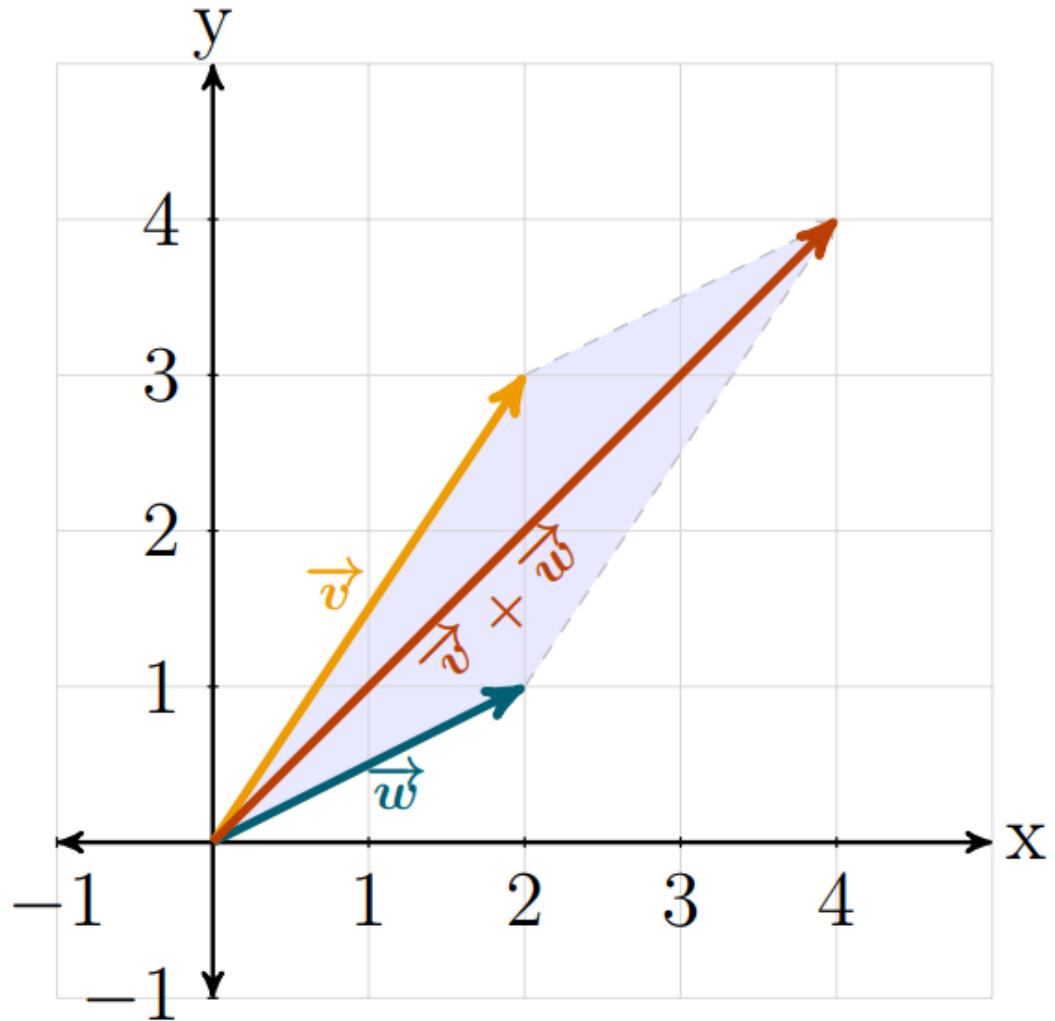
- Given two vector  $\vec{v}$  and  $\vec{w}$ , their sum,  $\vec{v} + \vec{w}$  is translation from the edge of  $\vec{v}$  with the magnitude and direction of  $\vec{w}$ .



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

$$\vec{w} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

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



$$\vec{z} = \vec{v} + \vec{w}$$

$$\vec{z} = \begin{pmatrix} 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$\vec{z} = \begin{pmatrix} 2+2 \\ 3+1 \end{pmatrix} = \begin{pmatrix} 4 \\ 4 \end{pmatrix}$$

# Scalar Multiplication

□ A scalar can be any real number.

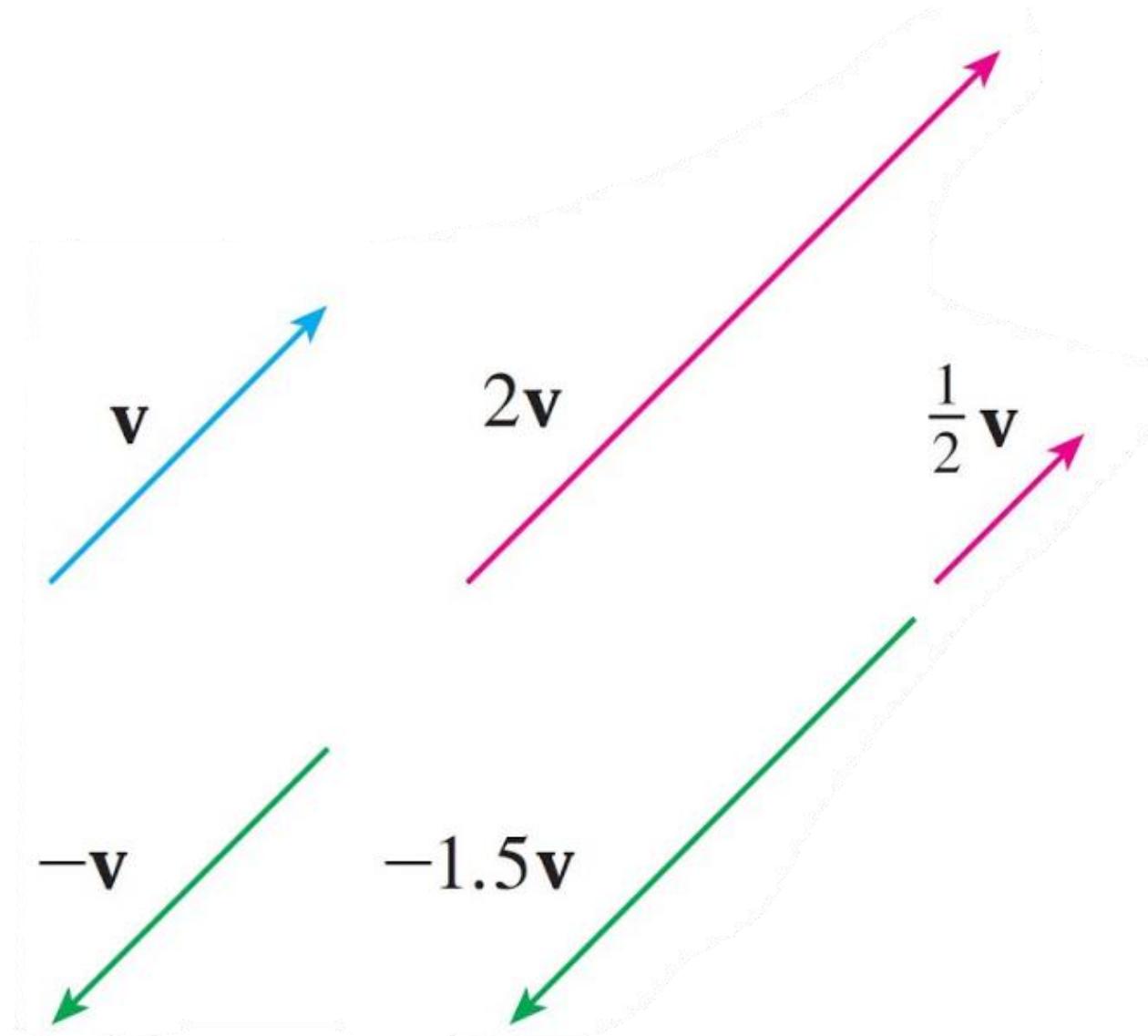


□ A real number is any number you can think of.

$$\lambda \in \mathbb{R}$$

There are four different outcomes when multiplying a vector by a scalar.

1. If  $\lambda > 1$  the vector will keep the same direction but stretch
2. If  $1 > \lambda > 0$  the vector will keep the same direction but shrink.
3. If  $\lambda < -1$  the vector will change direction and stretch.
4. If  $0 > \lambda > -1$  the vector will change direction and shrink.



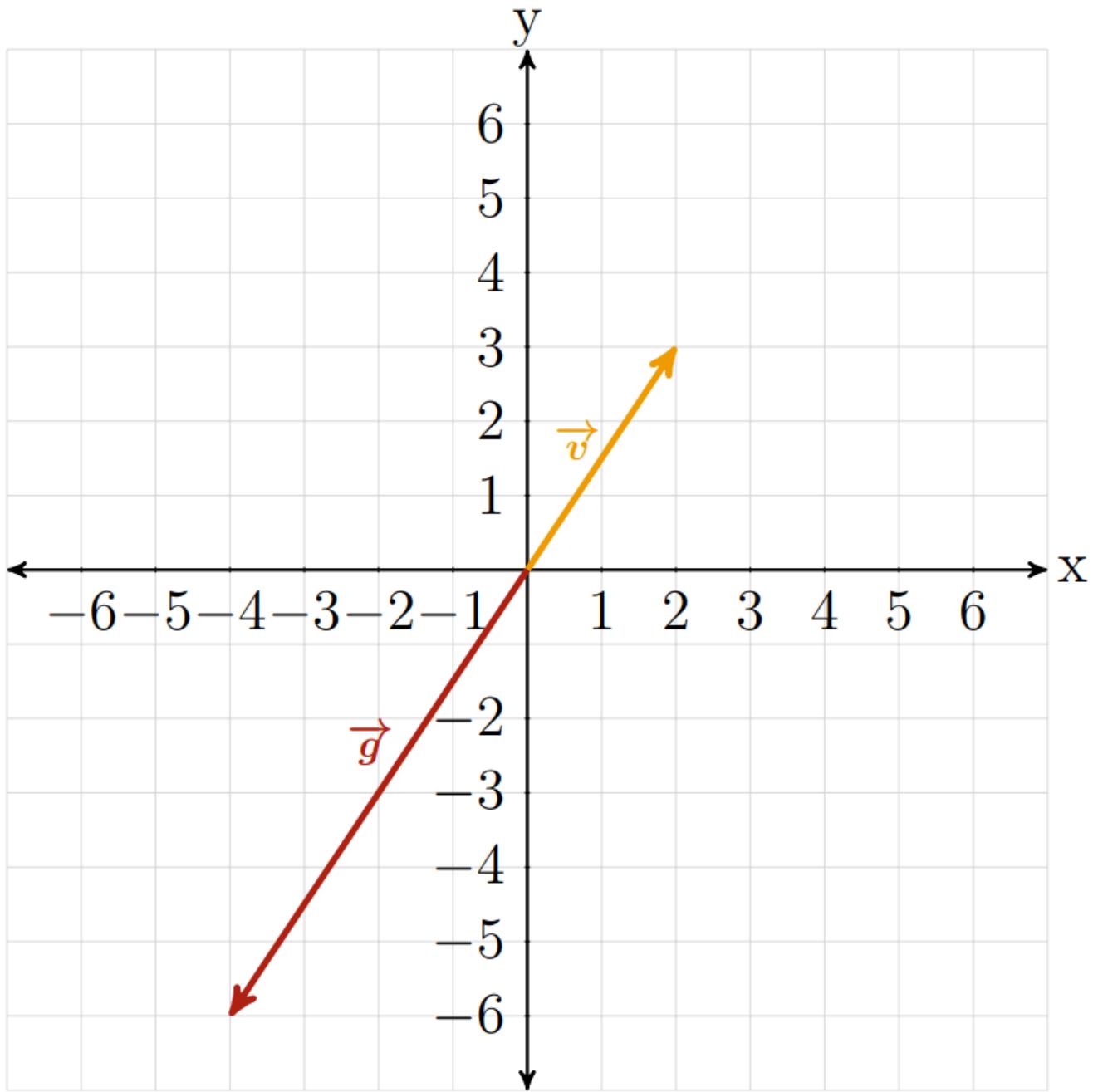
□ Multiplying a vector  $\vec{v}$  by a scalar  $\lambda$  can be defined by:  $\lambda\vec{v}$

□ If we define a new vector  $\vec{g}$  such that:

$$\vec{g} = \lambda\vec{v} \text{ and } \lambda = -2$$

$$\vec{g} = -2 \cdot \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$

$$\vec{g} = \begin{pmatrix} -4 \\ -6 \end{pmatrix}$$



# The Vector Space

- A **vector space** is a fundamental concept that provides a formal framework for the study and manipulation of vectors.
- Vectors are entities can be **added** together and **multiplied** by scalars.

- ❑ A vector space is a structured set consisting of a carefully defined group of entities, known as vectors, which adhere to specific rules.
- ❑ Vector space is describe using:  $(O, +, \times)$

- $O$  is an non-empty set whose elements are called vectors.
- $+$  is a binary operation (vector addition) that takes two vectors from  $O$  and produces another vector in  $O$ .
- $\times$  is an operation (scalar multiplication) that takes a scalar and a vector from  $O$  and produces another vector in  $O$

- If  $\vec{v}$  and  $\vec{w}$  are two vectors  $\in O$  then  $\vec{v} + \vec{w}$  must be in  $O$
- $\lambda \in \mathbb{R}$  and  $\vec{v} \in O$  then  $\lambda\vec{v}$  also needs to be in  $O$ .

# Axioms of Vector Space

- Commutative property of addition:

$$\vec{v} + \vec{w} = \vec{w} + \vec{v}$$

- Associative property of addition:

$$\vec{v} + (\vec{w} + \vec{t}) = (\vec{v} + \vec{w}) + \vec{t}$$

- Associative property of addition:

$$\vec{v} + 0 = \vec{v}$$

- An inverse element exists:

$$\vec{v} + (-\vec{v}) = \vec{0}$$

- Scalars can be distributed across the members of an addition:

$$c(\vec{v} + \vec{w}) = c\vec{v} + c\vec{w}$$

- Vector can be distributed to an addition of two scalars:

$$(c + d)\vec{v} = c\vec{v} + d\vec{v}$$

- The product of two scalars and an element is equivalent to one of the scalars being multiplied by the product of the other scalar and the element:

$$(cd)\vec{v} = c(d\vec{v})$$

- Multiplying an element by 1 just returns the same element:

$$1 \cdot \vec{v} = \vec{v}$$

# A Two-dimension Vector Space ( $\mathbb{R}^2$ )

- This is the space formed by all of the vectors with two dimensions, whose elements are real numbers.
- Let's define two vectors  $\vec{v}$  and  $\vec{w}$  as being in  $\mathbb{R}^2$ .
- Let's also define a scalar  $\lambda \in \mathbb{R}$

□ If we multiply  $\vec{v}$  by  $\lambda$  we have:

$$\lambda \cdot \vec{v} = \begin{pmatrix} \lambda \cdot v_1 \\ \lambda \cdot v_2 \end{pmatrix}$$

□ Multiplying a real number by a real number results in a new real number.

□ If we add  $\vec{v}$  and  $\vec{w}$ :

$$\vec{v} + \vec{w} = \begin{pmatrix} v_1 + w_1 \\ v_2 + w_2 \end{pmatrix}$$

# Subspaces

- The most important subsets of  $\mathbb{R}^n$  are those that closed under the operations of vector addition and scalar multiplication.
- They serve to generalize the geometric notions of points, line, and plane in two- and three-dimensional space.

□ A subspace of  $\mathbb{R}^n$  is a nonempty subset  $\emptyset \neq V \subseteq \mathbb{R}^n$  that satisfies.

- a) for every  $v, w \in V$ , the sum  $v + w \in V$  and
- b) for every  $v \in V$  and every  $c \in \mathbb{R}$ , the scalar product  $cv \in V$

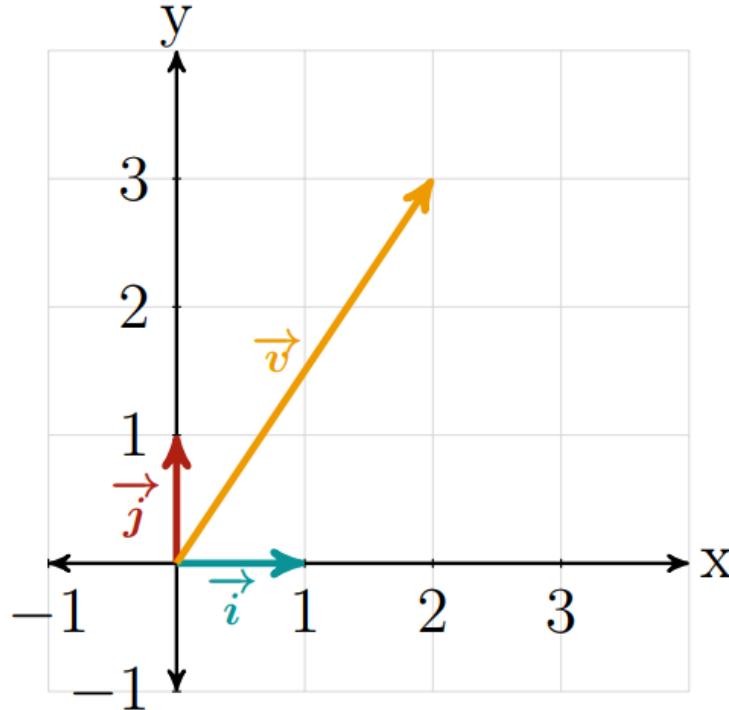
# Span and Linear Independence

□ Linear means there are no curves, just *lines*, *planes*, or *hyper-planes*, depending on the dimensions that we are working in.

# Example of linear combination

$$\alpha \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \beta \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad \text{where } \alpha, \beta \in \mathbb{R}$$

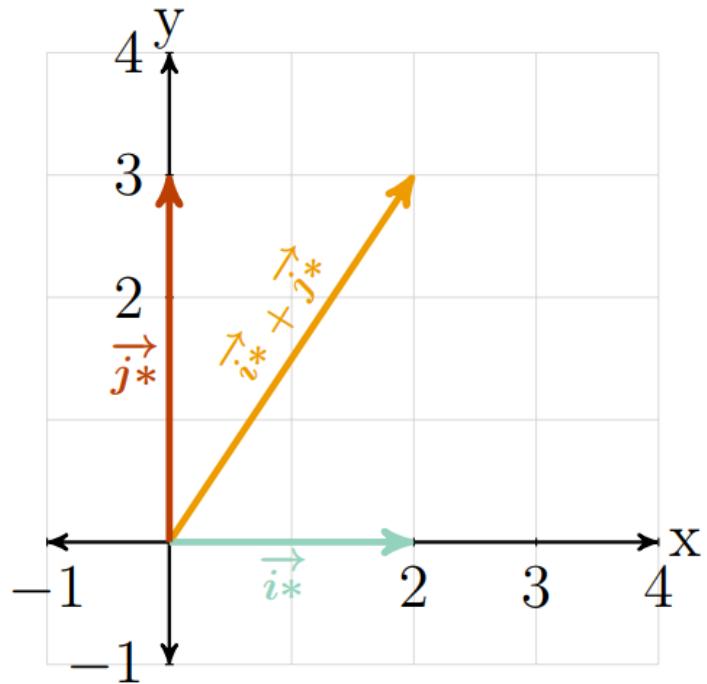
We name  $(1 \ 0)^T$  as  $\vec{i}$  and  $(0 \ 1)^T$  as  $\vec{j}$



$$\vec{v} = (2, 3)^T$$

We can stretch  $\vec{i}$  by **two** units and then sum it to a **three** units stretched version of  $\vec{j}$ , the result will be equal to  $\vec{v}$ .

$$\vec{v} = 2 \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 3 \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$



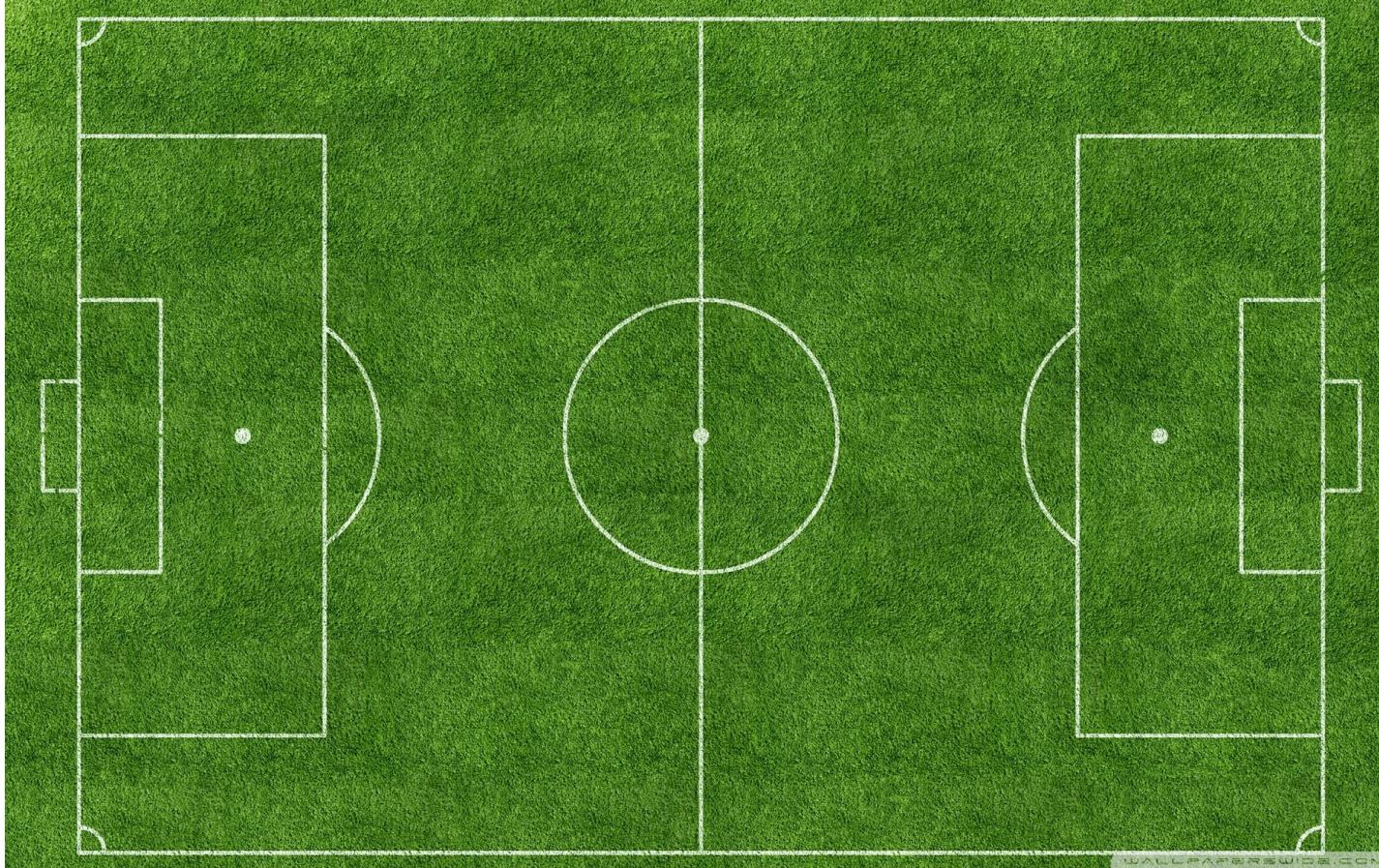
$$\vec{v} = 2 \cdot \vec{i} + 3 \cdot \vec{j}$$

The scalars 2 and 3 can be replaced with  $\alpha$  and  $\beta$ , where both of them are in  $\mathbb{R}$

$$\alpha \cdot \vec{i} + \beta \cdot \vec{j}$$

- We can display all the vectors of the vector space  $\mathbb{R}^2$  using  $\alpha \cdot \vec{i} + \beta \cdot \vec{j}$
- If I have the entire set of real numbers assigned to the scalars  $\alpha$  and  $\beta$ , it means that if I add up the scaled version of  $\vec{i}$  and  $\vec{j}$ , I can get any vector within  $\mathbb{R}^2$

- The vectors  $\vec{i}$  and  $\vec{j}$  are linearly independent.
- You can't get to  $\vec{j}$  via  $\vec{i}$  and vice versa.



- Linear independent can be defined using the equation:  $c_1 \cdot \vec{v}_1 + c_2 \cdot \vec{v}_3 + \dots + c_n \cdot \vec{v}_n = 0$
- The factors  $c_1, c_2, \dots, c_n$  are scalars or real numbers.
- The  $v$ 's are set of vectors that belong to the space are **linearly independent** if, and only if, the values for the  $c$ 's that satisfy that equality are 0.

$$\alpha \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \beta \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} = 0$$

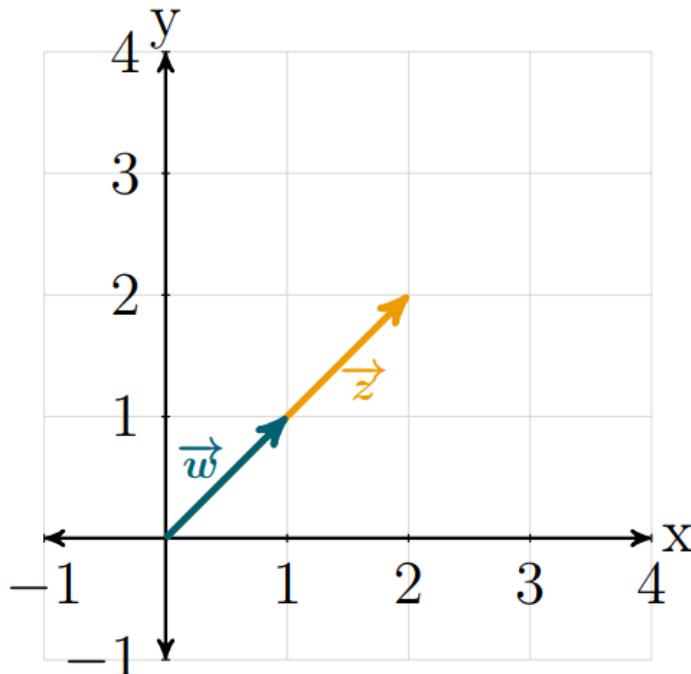
- The only way for the equality to be true is if both  $\alpha$  and  $\beta$  are equal to zero.
- Therefore,  $\vec{i}$  and  $\vec{j}$  are **linearly independent**.

□ The vectors  $v_1, \dots, v_k \in \mathbb{R}^n$  are called linearly dependent if there exist scalars  $c_1, \dots, c_k \in \mathbb{R}$ , not all zero, such that:

$$c_1v_1 + \cdots + c_kv_k = 0$$

□ To check **linear independence**, one needs to show that the only linear combination that produces the zero vector  $c_1 = \dots = c_k = 0$  is the *one and only* solution to the vector equation.

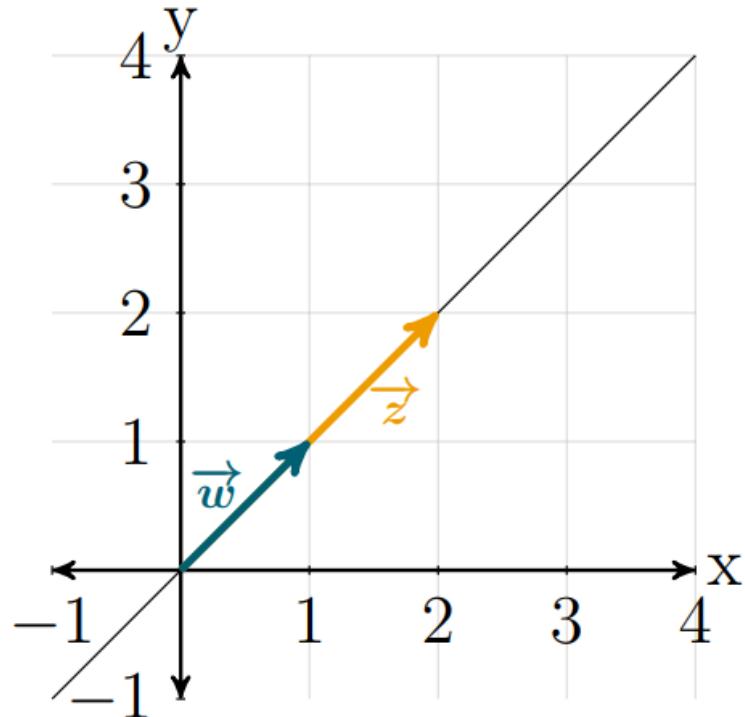
□ Let's consider  $\vec{w} = (1 \quad 1)^T$  and  $\vec{z} = (2 \quad 2)^T$



We can define a linear combination  
of these two vectors

$$\alpha \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} + \beta \cdot \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$

$$\vec{z} = 2 \cdot \vec{w}$$



We are not able to represent all of the vectors in the space using the two vectors,  $\vec{w}$  and  $\vec{z}$

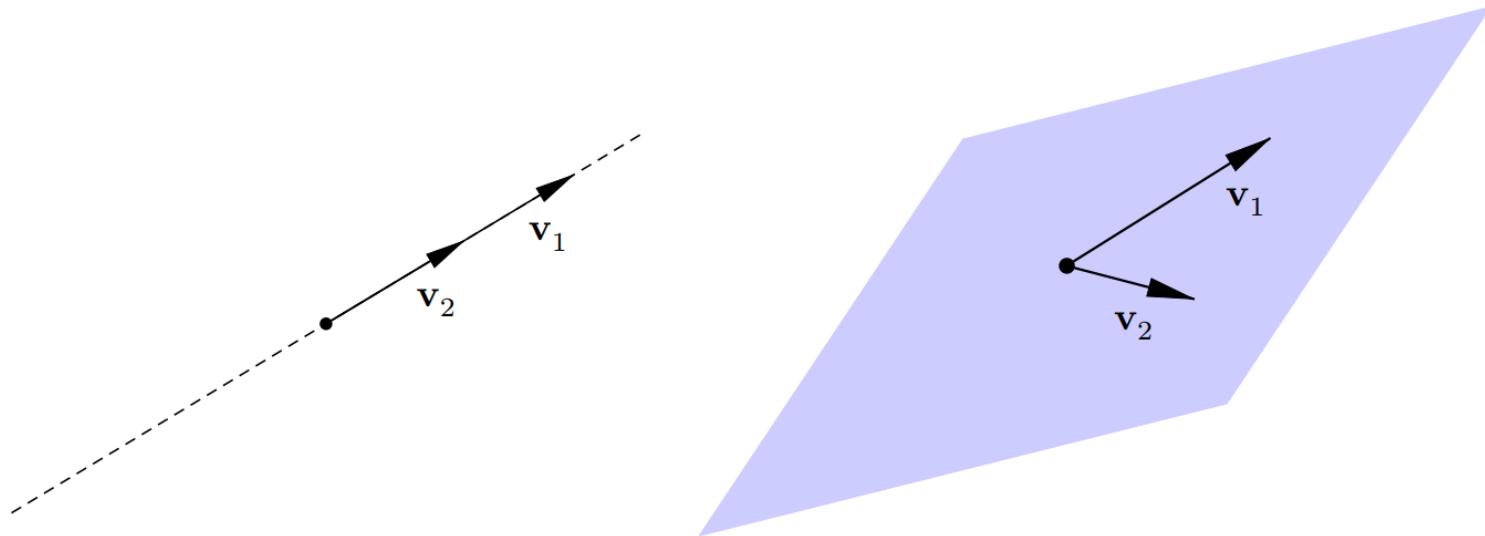
$\vec{z}$  and  $\vec{w}$  linearly dependence

□ Two vectors  $v, w \in V$  are linearly dependent if and only if they are parallel, meaning that one is a scalar multiple of the other.

- All the vectors that result from a linear combination define the **span**.
- The *span* of a finite collection of vectors  $v_1, \dots, v_k \in \mathbb{R}^n$  is the subsect  $V = \text{span}\{v_1, \dots, v_k\} \subset \mathbb{R}^n$  consisting of all possible linear combinations for  $c_1, \dots, c_k \in \mathbb{R}$

□  $\vec{i}$  and  $\vec{j}$  span the entire vector space (plane) because we can get all the vectors.

□ The span of  $\vec{w}$  and  $\vec{z}$  is a line.

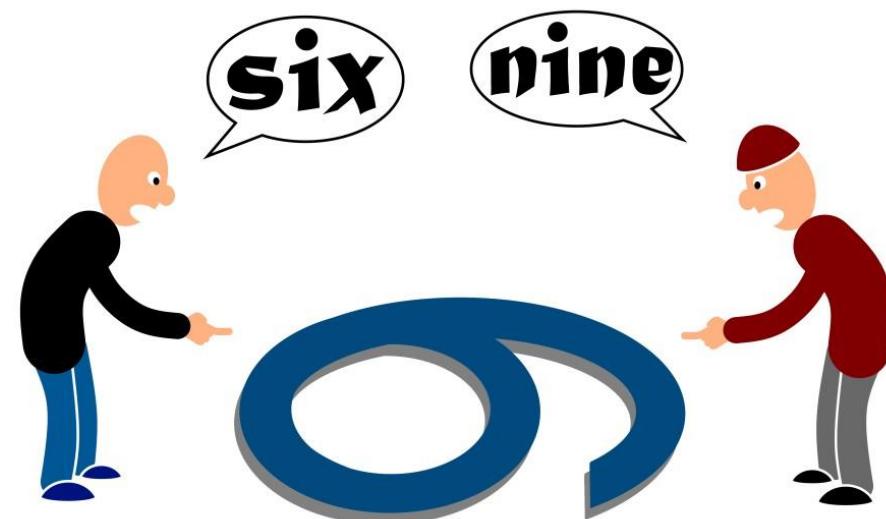
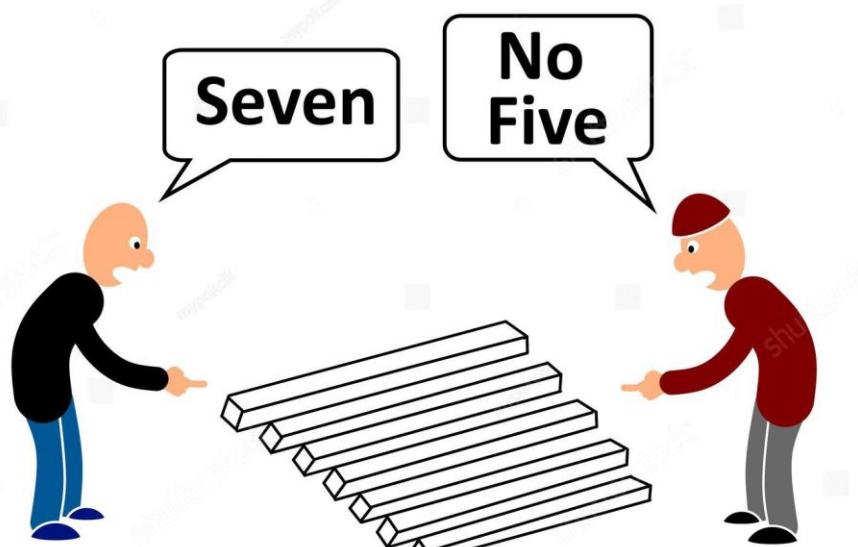


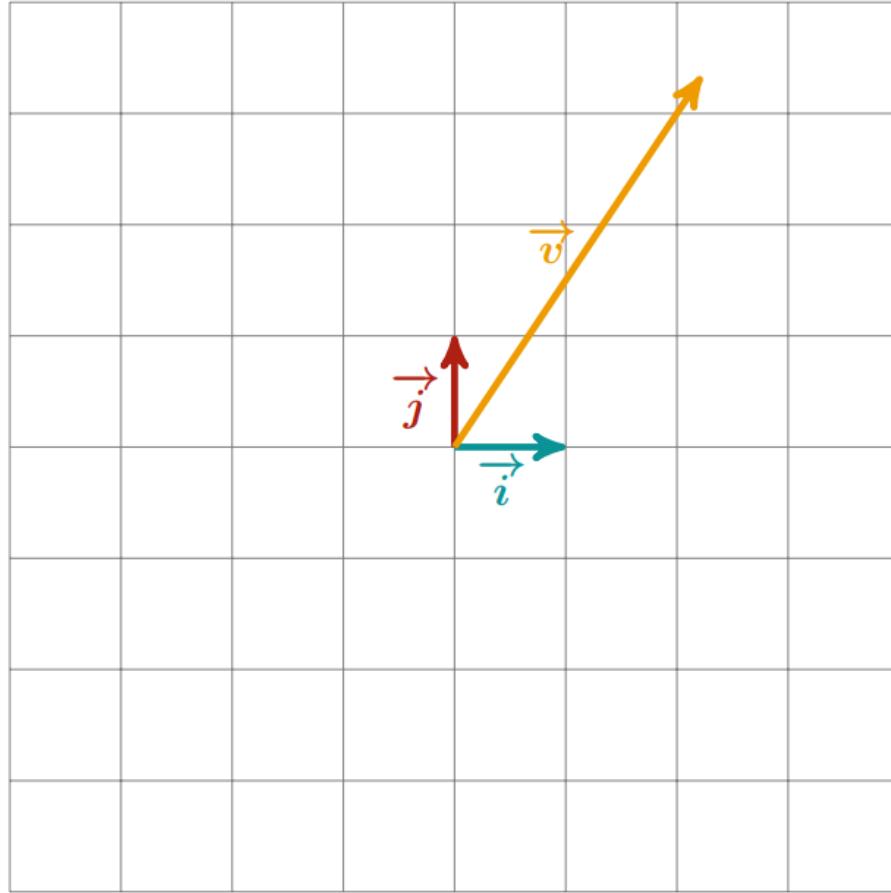
# Basis and Dimension

- A *basis* of a subspace  $V \subseteq \mathbb{R}^n$  is a finite set of vectors  $v_1, \dots, v_k \in V$  that span
  - (a) spans  $V$
  - (b) is linearly independent

- For a set of vectors to be considered a basis of a vector space, these vectors need to be linearly independent, and their span has to be equal to the entire space.
- $\vec{i}$  and  $\vec{j}$  form a basis of  $\mathbb{R}^2$ .

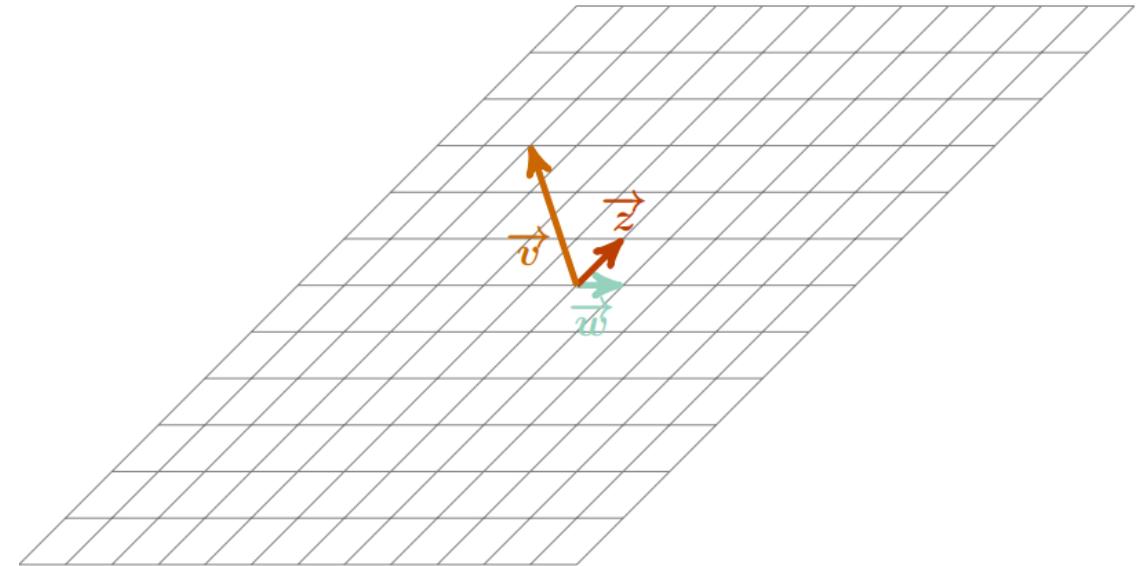
- A vector space can have more than one basis.
- Different basis is the **perspective** from which we observe the same vector in different ways.





$$\begin{aligned}\vec{i} &= (1 \quad 0)^T & \vec{j} &= (0 \quad 1)^T \\ \vec{v} &= (2 \quad 3)^T \\ \vec{v} &= 2 \cdot \vec{i} + 3 \cdot \vec{j}\end{aligned}$$

The grids represent our perspective or the basis, which is the way we observe  $\vec{v}$  on the basis formed by  $\vec{i}$  and  $\vec{j}$ .



Let's calculate the new coordinates of  $\vec{v}$  using the new basis above.

$$\vec{w} = (1 \quad 0)^T \quad \vec{z} = (1 \quad 1)^T$$

$$\vec{v} = (2 \quad 3)^T$$

$$\vec{v} = v_1^* \cdot \vec{w} + v_2^* \cdot \vec{z}$$

$$\begin{pmatrix} 2 \\ 3 \end{pmatrix} = v_1^* \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} + v_2^* \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Find  $v_1^*$  and  $v_2^*$

- The coordinates of  $\vec{v}$  in the new basis formed by  $\vec{w}$  and  $\vec{z}$  are  $(-1, 3)^T$

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{v}_2 = \begin{pmatrix} 1 \\ 1 \\ -1 \\ -1 \end{pmatrix}, \quad \mathbf{v}_3 = \begin{pmatrix} 1 \\ -1 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{v}_4 = \begin{pmatrix} 0 \\ 0 \\ 1 \\ -1 \end{pmatrix},$$

Wavelet basis plays an increasingly central role in modern signal and digital image processing.

Change of basis is very useful because you can find **properties** of vectors or data using a different basis or spaces, which allows for **faster** computation or even better results when dealing with machine learning models.

# Application

We wish to predict the price of a house



- The dataset consists of two measurements: the **number of bedrooms** and the **number of bathrooms**.
- The vector  $\vec{i} = (1, 0)^T$  will point in the directions where the number of bedrooms increases.
- The vector  $\vec{j} = (0, 1)^T$  will point in the directions where the number of bathrooms increases.

Let's assume that:

- ❑ Houses with more rooms in total tend to have higher prices.
- ❑ When the number of bathrooms is the same or close to the number of bedrooms, there is increase in price.

Let's define new basis:

□  $\vec{i}^* = (1, 1)^T$  and  $\vec{j}^* = (1, -1)^T$

- The vector  $\vec{i}^*$  will represent the total number of rooms.
- The vector  $\vec{j}^*$  displays the difference in number between bedrooms and bathrooms.

- We have a house with three bedrooms and two bathrooms.

$$\vec{w} = \begin{pmatrix} 3 \\ 2 \end{pmatrix}$$

- Let's check it from a new perspective, the one described by the new basis,  $\vec{i}^*$  and  $\vec{j}^*$

$$\begin{pmatrix} 3 \\ 2 \end{pmatrix} = w_1^* \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} + w_2^* \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Find the values of  $w_1^*$  and  $w_2^*$

**Thank you**

