



# Data 345

Applied Linear Algebra for Statistical Learning

Class 3 (Sep. 4, 2025)

# The Story So Far

- Created a basic <vector> object in Python.
- Use <\_\_repr\_\_> method to represent the object as an output.
- Define <\_\_add\_\_>, <\_\_mul\_\_>, and <\_\_rmul\_\_> methods to implement addition and scalar multiplication.

```
class vector:
    def __init__(self, data):
        self.data = data

    def __repr__(self):
        return f"vector({self.data})"

    def __add__(self, other):
        return vector(data=[a + b for a,b in zip(self.data, other.data)])

    def __mul__(self, other):
        return vector(data=[a*other for a in self.data])

    def __rmul__(self, other):
        return self.__mul__(other)
```

`vector([-1,3,9]) + -25*vector([44, 8, 3])`

✓ 0.0s

Python

`vector([-1101, -197, -66])`

✓ 0.0s

Python

# While You Were Away...

Converts input numbers to  
<float> type for computation

```
class vector:
    def __init__(self, data):
        self.data = data
```

```
class vector:
    def __init__(self, data):
        try:
            # Convert input data to a float type for computation
            self.data = [float(x) for x in data]
        except (TypeError, ValueError) as e:
            raise type(e)(f"Input must be an iterable of numbers. Problem: {e}") from None

        # Insist that the list of elements is not empty
        if len(self.data) == 0:
            raise ValueError("Vector cannot be empty")
```

Basic input validation. Returns an error when:

- The input is not an iterable (list-like)
- The input is an iterable, but contains things that can't be converted to numbers
- The input list is empty

```
ValueError: Input must be an iterable of numbers. Problem: could not convert string to float: 'blah' Cell Execution Error
```

```
vector([1, 2, 'blah'])
```

```
TypeError: Input must be an iterable of numbers. Problem: 'int' object is not iterable Cell Execution Error
```

```
vector(15345)
```

```
ValueError: Vector cannot be empty Cell Execution Error
```

```
class vector(data: list[Any])
```

```
vector([])
```


# While You Were Away...

More error handling –  
make sure we do not  
add non-vectors to  
vectors, or vectors of  
different sizes

```
def __add__(self, other):  
    return vector(data=[a + b for a,b in zip(self.data, other.data)])  
  
def __mul__(self, other):  
    return vector(data=[a*other for a in self.data])  
  
def __rmul__(self, other):  
    return self.__mul__(other)
```

For scalar  
multiplication,  
we should only  
use scalars.

```
def __add__(self, other):  
    # Check that we are adding vectors to vectors  
    if not isinstance(other, vector):  
        raise TypeError(f"Can only add vector to vector, got {type(other).__name__}")  
  
    # Check that the vectors are the same length  
    if len(self.data) != len(other.data):  
        raise ValueError(f"Vector dimensions must match: {len(self.data)} vs {len(other.data)}")  
  
    return vector(data=[a + b for a,b in zip(self.data, other.data)])  
  
def __mul__(self, other):  
    # Check that we are multiplying by a scalar.  
    if not isinstance(other, (int, float)):  
        raise TypeError(f"Cannot multiply vector by {type(other).__name__}")  
    return vector(data=[a*other for a in self.data])  
  
def __rmul__(self, other):  
    return self*other
```



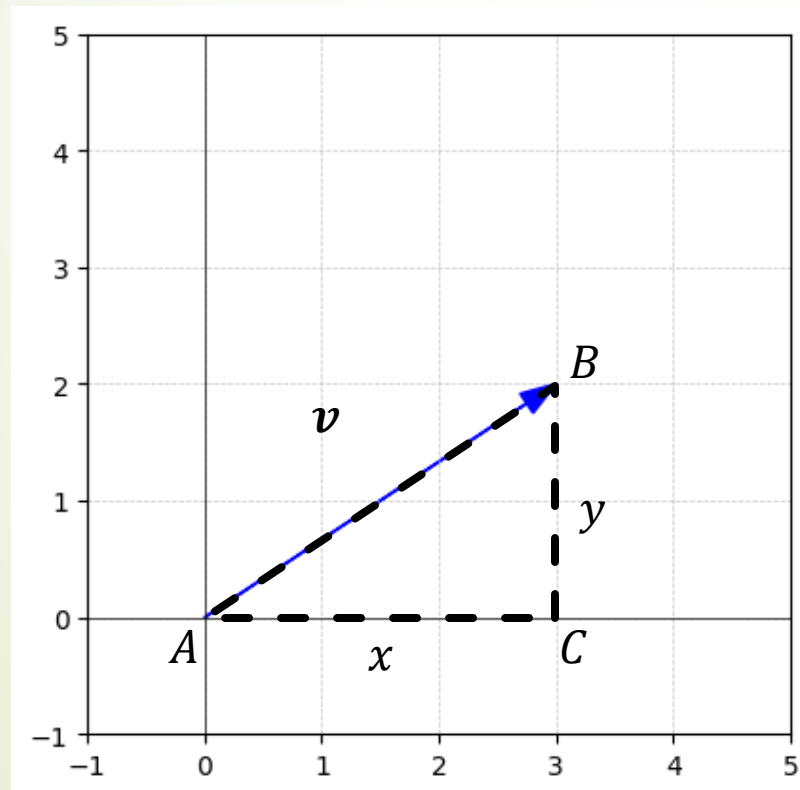
# What's Next?

- Implement vector subtraction. If  $v = [v_1, v_2, \dots, v_n]$  and  $w = [w_1, w_2, \dots, w_n]$ , then  $v - w = [v_1 - w_1, v_2 - w_2, \dots, v_n - w_n]$ . This is easy to do by combining both addition and scalar multiplication.
- Implement vector length as well as vector angles.
- Cauchy-Schwarz and Triangle inequalities.
- Matrices: matrix addition, scalar multiplication, and matrix multiplication.
- Row operations for matrices.

# Vector Length

- Recall that we realized vectors geometrically as directed line segments emanating from the origin.
- The “length” of a vector, then, should correspond to the length of this line segment.

$$\mathbf{v} = [x, y]$$



$$\text{length}(\mathbf{v}) = \text{length}(\overline{AB})$$

$$\begin{aligned} \left(\text{length}(\overline{AB})\right)^2 &= \text{length}(\overline{AC})^2 + \text{length}(\overline{BC})^2 \\ &= x^2 + y^2 \end{aligned}$$

Therefore,

$$\text{length}(\mathbf{v}) = \sqrt{x^2 + y^2}$$

# Vector Length

- Let  $\mathbf{v} = [v_1, v_2, \dots, v_n]$ . Then the **Euclidean norm** of  $\mathbf{v}$ , written  $\|\mathbf{v}\|$ , is given by  $\|\mathbf{v}\| = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$ . This is the “usual” length assigned to a vector in  $\mathbb{R}^n$ .

$$\begin{array}{c} \mathbf{v} = [v_1, v_2, v_3, \dots, v_n] \\ \updownarrow \updownarrow \updownarrow \updownarrow \\ \mathbf{v} = [v_1, v_2, v_3, \dots, v_n] \end{array}$$

$$\|\mathbf{v}\|^2 = (v_1 \cdot v_1) + (v_2 \cdot v_2) + (v_3 \cdot v_3) + \dots + (v_n \cdot v_n)$$

$$\begin{array}{c} \mathbf{v} = [v_1, v_2, v_3, \dots, v_n] \\ \updownarrow \updownarrow \updownarrow \updownarrow \\ \mathbf{w} = [w_1, w_2, w_3, \dots, w_n] \end{array}$$

$$(v_1 \cdot w_1) + (v_2 \cdot w_2) + (v_3 \cdot w_3) + \dots + (v_n \cdot w_n)$$



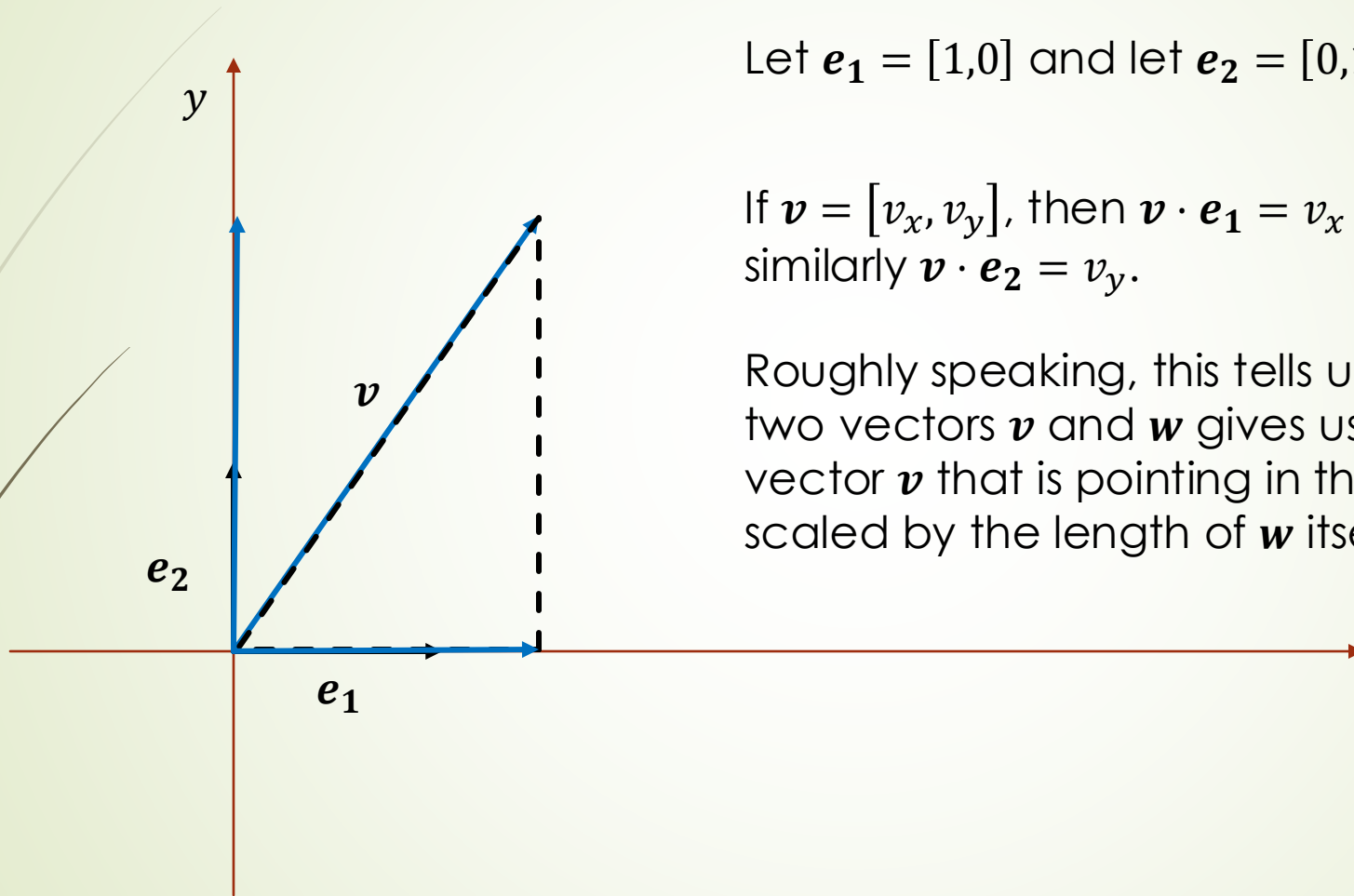


# Vector Dot Product

- ▶ Let  $\mathbf{v} = [v_1, v_2, \dots, v_n]$  and  $\mathbf{w} = [w_1, w_2, \dots, w_n]$  be two  $n$ -dimensional vectors. Then the **vector dot product** of  $\mathbf{v}$  and  $\mathbf{w}$  (or **scalar product** of  $\mathbf{v}$  and  $\mathbf{w}$ ), written as  $\mathbf{v} \cdot \mathbf{w}$  (or  $\langle \mathbf{v}, \mathbf{w} \rangle$ ), is given as  $\mathbf{v} \cdot \mathbf{w} = (v_1 w_1) + (v_2 w_2) + \dots + (v_n w_n)$ .
- ▶ **Note:** this is not multiplication for vectors. The output of the vector dot product is a scalar, not a vector.
- ▶ From our earlier definition of  $\|\mathbf{v}\|$ , it is easy to see that  $\mathbf{v} \cdot \mathbf{v} = \|\mathbf{v}\|^2$ . How should we interpret  $\mathbf{v} \cdot \mathbf{w}$ ?



# Vector Dot Product



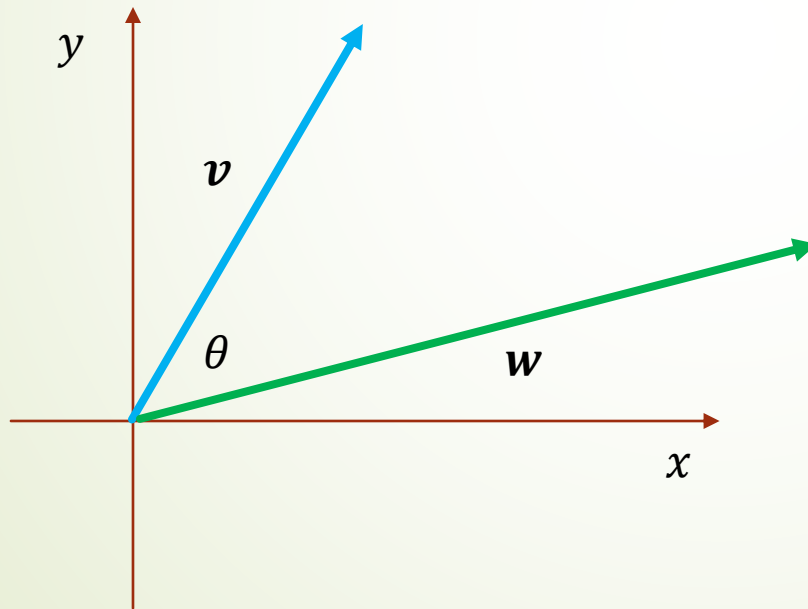
Let  $\mathbf{e}_1 = [1,0]$  and let  $\mathbf{e}_2 = [0,1]$ .

If  $\mathbf{v} = [v_x, v_y]$ , then  $\mathbf{v} \cdot \mathbf{e}_1 = v_x \cdot 1 + v_y \cdot 0 = v_x$ , and similarly  $\mathbf{v} \cdot \mathbf{e}_2 = v_y$ .

Roughly speaking, this tells us that the dot product of two vectors  $\mathbf{v}$  and  $\mathbf{w}$  gives us the length “part” of the vector  $\mathbf{v}$  that is pointing in the same direction as  $\mathbf{w}$ , but scaled by the length of  $\mathbf{w}$  itself.

# Vector Dot Product

- Geometrically, the dot product has an additional useful realization:



$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cdot \cos(\theta)$$

Could be negative

Always nonnegative

# Vector Dot Product

- Recall that we discussed the notion of **simultaneous linear equations** in multiple variables.

$$3x_1 + 2x_2 + x_3 = 39$$

$$2x_1 + 3x_2 + x_3 = 34$$

$$x_1 + 2x_2 + 3x_3 = 26$$

- Crucially, we can also set up this system of equations in terms of vector dot products.
- Setting  $\mathbf{v}_1 = [3, 2, 1]$ ,  $\mathbf{v}_2 = [2, 3, 1]$ ,  $\mathbf{v}_3 = [1, 2, 3]$ , and  $\mathbf{x} = [x_1, x_2, x_3]$ , the system of equations becomes:

$$\mathbf{v}_1 \cdot \mathbf{x} = 39$$

$$\mathbf{v}_2 \cdot \mathbf{x} = 34$$

$$\mathbf{v}_3 \cdot \mathbf{x} = 26$$



# Goals

- Implement a vector subtraction method
- Implement a vector length or `<norm>` attribute for the `<vector>` class (take a vector  $\mathbf{v} = [v_1, v_2, \dots, v_n]$  and output  $\sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$ )
- Implement a vector `<dot>` method, which will take in two like-sized vectors  $\mathbf{v} = [v_1, v_2, \dots, v_n]$  and  $\mathbf{w} = [w_1, w_2, \dots, w_n]$ , and then output  $v_1w_1 + v_2w_2 + \dots + v_nw_n$
- Utilize the properties of the vector dot product and a math computation library to calculate the angle between two vectors.