

# Linear Algebra Made Simple

A friendly introduction to scalars, vectors, matrices, and array operations with easy examples you can work through in your head.

# Understanding the Building Blocks

## Scalars

Just a single number on its own

$$a = 5$$

## Vectors

A list of numbers arranged in order

$$\mathbf{v} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

## Matrices

A rectangle filled with numbers in rows and columns

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

These three types form the foundation of linear algebra. Think of scalars as single values, vectors as ordered lists, and matrices as organised grids.



# Let's Try Vector Addition

## The Question

Given the vector:

$$\mathbf{v} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

Add 2 to every entry.

## The Answer

Simply add 2 to each number:

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

Each element increases by the same amount.

# Matrix Operations Are Similar

## Question

Take this matrix and add 1 to every entry:

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

## The Process

Add 1 to each number in the grid, maintaining the same structure and position.

## The Result

$$\begin{bmatrix} 1 + 1 & 2 + 1 \\ 3 + 1 & 4 + 1 \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ 4 & 5 \end{bmatrix}$$

# Understanding Shapes and Dimensions

01

## Reading Matrix Shape

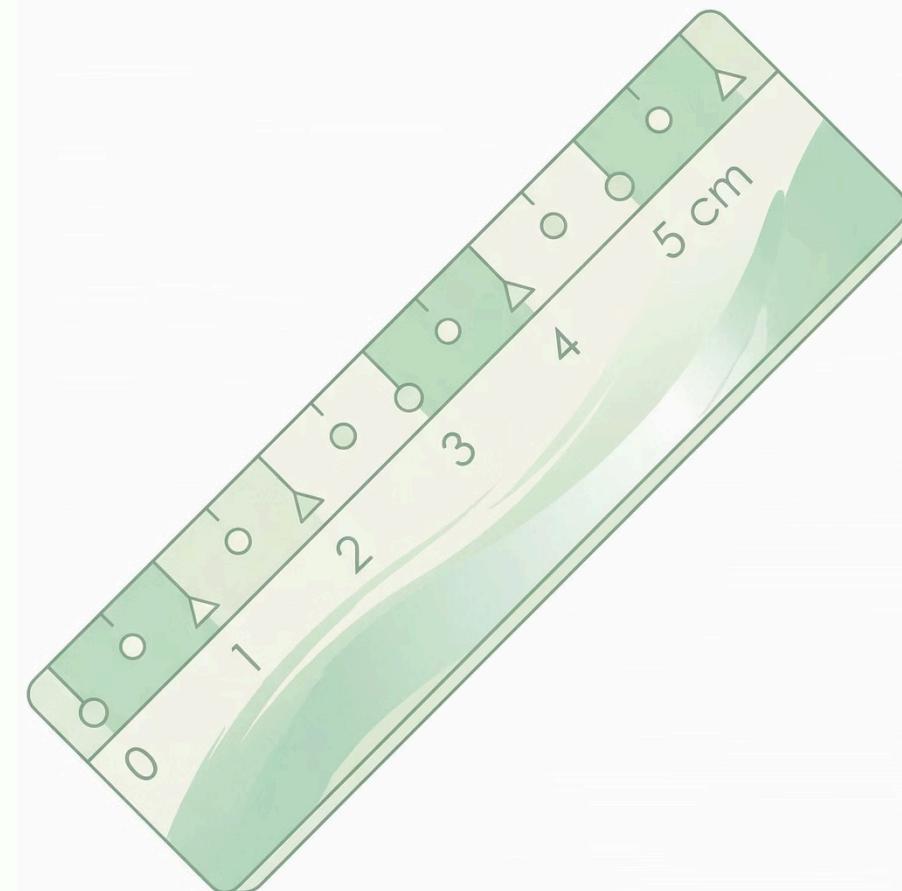
For matrix  $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$ , count rows first (2), then columns (3). The shape is  $2 \times 3$ .

02

## Multi-dimensional Shapes

A batch of 10 images, each  $28 \times 28$  pixels with 1 colour channel, has shape  $(10, 28, 28, 1)$ . The first number (10) represents the batch size.

- **Tip:** The shape tells you the dimensions at each level. Think of it like giving directions: batch → height → width → channels.



# Element-wise Operations

## Vector Addition

Given two vectors:

$$\mathbf{x} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}, \mathbf{y} = \begin{bmatrix} 4 \\ 5 \end{bmatrix}$$

Add them element by element:

$$\mathbf{x} + \mathbf{y} = \begin{bmatrix} 1 + 4 \\ 3 + 5 \end{bmatrix} = \begin{bmatrix} 5 \\ 8 \end{bmatrix}$$

## Scalar Multiplication

For matrix:

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

Multiply every entry by 2:

$$2A = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

Element-wise operations work on corresponding positions. Each element operates independently with its counterpart.



# CONNECTIONS

# The Dot Product Explained

- 1
- 2
- 3

**Start with Two Vectors**

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

**Multiply Pairs**

$$1 \cdot 3 = 3$$

$$2 \cdot 4 = 8$$

**Add Results**

$$3 + 8 = 11$$

That's your dot product!

The dot product combines two vectors into a single number by multiplying corresponding entries and summing them up.

# Matrix-Vector Multiplication

## Example Problem

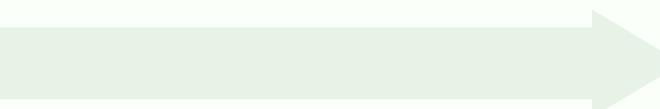
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, B = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Find  $AB$



### First Row Calculation

Dot product of first row with vector:  $1 \cdot 1 + 2 \cdot 1 = 3$



### Second Row Calculation

Dot product of second row with vector:  $3 \cdot 1 + 4 \cdot 1 = 7$



### Combine the Results

$$AB = \begin{bmatrix} 3 \\ 7 \end{bmatrix}$$

# Broadcasting: NumPy's Clever Trick

Broadcasting automatically expands smaller arrays to match larger ones, making operations more intuitive.

## Adding a Scalar

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + 10 = \begin{bmatrix} 11 \\ 12 \\ 13 \end{bmatrix}$$

The single number 10 "broadcasts" to match each element.

## Adding a Row Vector

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + [10 \quad 20]$$

Result:  $\begin{bmatrix} 11 & 22 \\ 13 & 24 \end{bmatrix}$

The row broadcasts across each matrix row.



# Reshaping and Transposing

## Reshaping Arrays

Take 6 numbers  $[1, 2, 3, 4, 5, 6]$  and reshape to  $2 \times 3$ :

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

The total count stays the same (6 elements), just arranged differently.

## The Transpose Operation

For matrix  $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$ , the transpose  $A^\top$  flips rows and columns:

$$A^\top = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

Row 1 becomes column 1, and row 2 becomes column 2.

These operations change how we view data without changing the underlying values. They're essential tools in linear algebra and machine learning.