

# Basic Linear Algebra

# What Is Linear Algebra?

**Linear algebra (in deep learning):**

A mathematical language to express many operations at once.

# Vector: An Array of Numbers

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

# Element-Wise Vector Operations

$$\mathbf{v} + \mathbf{w} = \begin{bmatrix} v_1 + w_1 \\ v_2 + w_2 \\ \vdots \\ v_n + w_n \end{bmatrix}$$

$$\mathbf{v} - \mathbf{w} = \begin{bmatrix} v_1 - w_1 \\ v_2 - w_2 \\ \vdots \\ v_n - w_n \end{bmatrix}$$

$$\mathbf{v} \cdot \mathbf{w} = \begin{bmatrix} v_1 \cdot w_1 \\ v_2 \cdot w_2 \\ \vdots \\ v_n \cdot w_n \end{bmatrix}$$

$$\mathbf{v} / \mathbf{w} = \begin{bmatrix} v_1 / w_1 \\ v_2 / w_2 \\ \vdots \\ v_n / w_n \end{bmatrix}$$

...

# Matrix: A 2D Array of Numbers

$$\mathbf{M} = \begin{bmatrix} M_{1,1} & M_{1,2} & \cdots & M_{1,m} \\ M_{2,1} & M_{2,2} & \cdots & M_{2,m} \\ \vdots & \vdots & \ddots & \vdots \\ M_{n,1} & M_{n,2} & \cdots & M_{n,m} \end{bmatrix}$$

# Matrix Transpose

$$\mathbf{M} = \begin{bmatrix} M_{1,1} & M_{1,2} & M_{1,3} & M_{1,4} \\ M_{2,1} & M_{2,2} & M_{2,3} & M_{2,4} \end{bmatrix} \rightarrow \mathbf{M}^T = \begin{bmatrix} M_{1,1} & M_{2,1} \\ M_{1,2} & M_{2,2} \\ M_{1,3} & M_{2,3} \\ M_{1,4} & M_{2,4} \end{bmatrix}$$

# Row Vector: The Transpose of a Vector

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \rightarrow \mathbf{v}^\top = [v_1, v_2, \dots, v_n]$$

# Matrix Multiplication

$$C_{i,j} = \sum_{k=1}^p A_{i,k} \cdot B_{k,j}$$



# Matrix Multiplication: Example

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \\ b_{41} & b_{42} & b_{43} \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \end{bmatrix}$$

$2 \times 4$

$4 \times 3$

$2 \times 3$

# Vector Multiplication

A *column vector* is multiplied on the right:

$$\mathbf{M}\mathbf{v}$$

A *row vector* is multiplied on the left:

$$\mathbf{v}^T \mathbf{M}$$

# Inner product (dot product)

$$\mathbf{v}^\top \mathbf{w} = \sum_{i=1}^n v_i w_i$$

# Outer Product

$$M = \mathbf{vw}^T$$

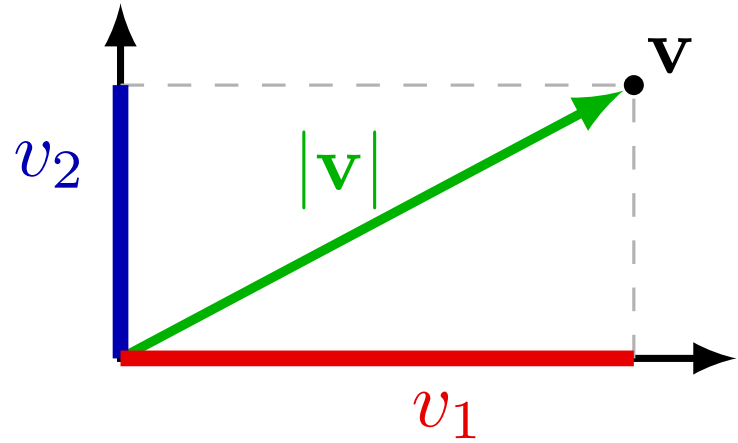
# Euclidean Norm: Length of a Vector

$$\|\mathbf{v}\| = \sqrt{\mathbf{v}^\top \mathbf{v}} = \sqrt{\sum_{i=1}^n v_i^2}$$

# Euclidean Norm: 2D Example

Length of  $\mathbf{v}$ :

$$\|\mathbf{v}\| = \sqrt{v_1^2 + v_2^2}$$



# Frobenius Norm: Magnitude of a Matrix

$$\|\mathbf{M}\| = \sqrt{\sum_{i=1}^n \sum_{j=1}^m M_{i,j}^2}$$

# Linear Algebra Through an Example

Given  $a_1, a_2, \dots, a_n$  compute:

$$b_i = a_i - \mu_a$$

where  $\mu_a = \frac{1}{n} \sum_{i=1}^n a_i$

Without linear algebra:

```
# Assume a = [a_1, a_2, ...] is given
S = 0
for v in a:
    S += v
mean = S / len(a)
b = []
for v in a:
    b.append(v - mean)
```

With linear algebra:

```
b = a - a.dot(torch.ones_like(a)) / len(a)
```

With pytorch:

```
b = a - a.mean()
```



# Basic Linear Algebra - TL;DR

Linear algebra: a mathematical language to express many operations at once