## Introduction to Linear Algebra

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#### Introduction

#### Linear algebra deals with:

- ► Vectors.
- ▶ Matrices. (The linear transformations of vector spaces.)

#### It is used in many places:

- ▶ Physics (velocity, force, ...).
- ► Artificial intelligence neural networks.
- ► Adaptive Law-Based Transformation (ALT).
- **.**..

### What is a Vector?

- ▶ Mathematician's view: A set that satisfies the 8 axiom.
- ▶ Physicist's view: An arrow in space with direction and magnitude.
- ► Computer Scientist's view: A 1D array or list of numbers.

$$\vec{v} = \begin{bmatrix} 3 \\ -1 \\ 2 \end{bmatrix}, \quad \mathbf{w} = \langle 4, 0, -5 \rangle, \quad \vec{u} = (1, 2, 3).$$
 (1)

- ► Notation:
  - ▶ Boldface: v (common in CS and mathematical books)
  - ightharpoonup Arrow:  $\vec{v}$  (common in physics)
  - ► Angled brackets or parentheses

## Example 1 (Vectors)

$$\vec{a} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \vec{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad \vec{c} = \begin{bmatrix} -3 \\ 2 \\ 5 \end{bmatrix}, \quad \vec{0} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$
 (2)



## What is the Dot Product?

▶ The dot product of two vectors  $\vec{a}$  and  $\vec{b}$  is:

$$\vec{a} \cdot \vec{b} = a_1 b_1 + a_2 b_2 + \dots + a_n b_n.$$
 (3)

▶ It can be shown that:

$$\vec{a} \cdot \vec{b} = ||\vec{a}|| \cdot ||\vec{b}|| \cos(\angle(\vec{a}, \vec{b})).$$
 (4)

## Example 2 (Dot product)

$$\begin{bmatrix} 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ -1 \end{bmatrix} = 2 \cdot 4 + 3 \cdot (-1) = 8 - 3 = 5.$$
 (5)



## Dot Product - Practice Exercises

► Calculate the dot product of the following vector pairs:

$$1. \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ 0 \\ -1 \end{bmatrix}.$$

$$2. \begin{bmatrix} 2 \\ 5 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

## Dot Product – Solutions

▶ Solution to (a):

$$1 \cdot 4 + (-2) \cdot 0 + 3 \cdot (-1) = 4 + 0 - 3 = \boxed{1}.$$

▶ Solution to (b):

$$2 \cdot (-1) + 5 \cdot 1 = -2 + 5 = \boxed{3}.$$

- ► Interpretation:
  - ightharpoonup (a) Zero  $\Rightarrow$  the vectors are perpendicular.
  - ightharpoonup (b) Positive  $\Rightarrow$  angle between the vectors is acute.



### What is a Matrix?

▶ A 2D array of numbers arranged in rows and columns:

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

where A is a  $2 \times 3$  matrix.

- ► Types:
  - Square: same number of rows and columns (e.g.,  $2 \times 2$ ,  $3 \times 3$ ). Mathematicians like them.
  - Non-square: different number of rows and columns (e.g.,  $2 \times 3$ ,  $3 \times 2$ ).



# Vector-Matrix Multiplication

► Each row of the matrix is dotted with the vector:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}, \quad \vec{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}. \tag{7}$$

► Multiplication:

$$A\vec{v} = \begin{bmatrix} 1 \cdot 2 + 2 \cdot 1 \\ 3 \cdot 2 + 4 \cdot 1 \\ 5 \cdot 2 + 6 \cdot 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 10 \\ 16 \end{bmatrix}. \tag{8}$$

## Vector-Matrix Multiplication – Exercises

- ▶ Multiply the vector  $\vec{v}$  with matrix M in the following examples:
  - 1. General case:

$$\vec{v} = \begin{bmatrix} 1\\2 \end{bmatrix}, \quad M = \begin{bmatrix} 3 & 0 & 1\\ 1 & 4 & 2 \end{bmatrix}. \tag{9}$$

2. Interesting effect:

$$\vec{v} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad M_1 = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad M_2 = \begin{bmatrix} -2 & -2 \\ -2 & -2 \end{bmatrix}.$$
 (10)

3. Even more interesting effect:

$$\vec{v} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad M_3 = \begin{bmatrix} 1 & -1 \\ 2 & -2 \end{bmatrix}, \quad M_4 = \begin{bmatrix} 5 & -5 \\ -3 & 3 \end{bmatrix}.$$
 (11)

► Try to interpret the geometric meaning of each result.



# Vector-Matrix Multiplication – Solutions

1. General case:

$$\vec{v} \cdot M = \begin{bmatrix} 1\\2 \end{bmatrix} \cdot \begin{bmatrix} 3 & 0 & 1\\1 & 4 & 2 \end{bmatrix} = \begin{bmatrix} 5\\8\\5 \end{bmatrix}. \tag{12}$$

2. Stretching cases:

$$\vec{v} \cdot M_1 = \begin{bmatrix} 2\\2 \end{bmatrix} = 2\vec{v}, \quad \vec{v} \cdot M_2 = \begin{bmatrix} -4\\-4 \end{bmatrix} = -4\vec{v}.$$
 (13)

3. Nearly null output:

$$\vec{v} \cdot M_3 = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = 0\vec{v}, \quad \vec{v} \cdot M_4 = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = 0\vec{v}.$$
 (14)



## Matrices as linear transformation

- ▶ Each matrix denotes a transformation  $\vec{v} \mapsto M\vec{v}$ . It's always linear.
- ▶ It can be proven that all linear transformation can be described with a matrix.

## Transformation examples

- Stretching:  $\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ .
- Rotation:  $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ .
- $\blacktriangleright \text{ Reflection: } \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}.$
- ▶ Projection (flattening onto a line or plane)  $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ .



## Matrix-Matrix Multiplication

Procedure is the same as the Vector-Matrix multiplication. If an  $n \times m$  matrix is multiplied by an  $m \times k$ , the result is an  $n \times k$  matrix.

▶ The number of columns in the first matrix must be equal to the number of rows in the second matrix.

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$
 (15)

$$A \cdot B = \begin{bmatrix} a_{11} \cdot b_{11} + a_{12} \cdot b_{21} & a_{11} \cdot b_{12} + a_{12} \cdot b_{22} \\ a_{21} \cdot b_{11} + a_{22} \cdot b_{21} & a_{21} \cdot b_{12} + a_{22} \cdot b_{22} \end{bmatrix}$$
(16)

## Example 3

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \quad B = \begin{bmatrix} 2 & 0 \\ 1 & 2 \end{bmatrix} \Rightarrow A \cdot B = \begin{bmatrix} 4 & 4 \\ 10 & 8 \end{bmatrix}$$
 (17)



## Other examples

## Example 4

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow A \cdot B = \begin{bmatrix} 4 & 5 \\ 10 & 11 \end{bmatrix}$$
 (18)

## Example 5

$$A = \begin{bmatrix} 8 & 4 & 11 & 5 \\ 4 & 4 & 21 & 6 \\ 6 & 9 & 11 & 8 \end{bmatrix}, B = \begin{bmatrix} 3 & 2 \\ 4 & 6 \\ 1 & 4 \\ 7 & 2 \end{bmatrix} \Rightarrow A \cdot B = \begin{bmatrix} 86 & 94 \\ 91 & 128 \\ 121 & 126 \end{bmatrix}$$
 (19)

# Eigenvalues and Eigenvectors

► Notice that

$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}. \tag{20}$$

ightharpoonup Generally for a square matrix A if

$$A\vec{v} = \lambda \vec{v},\tag{21}$$

#### then:

- $\vec{v}$  is an eigenvector and
- $\triangleright$   $\lambda$  is the corresponding eigenvalue.
- $\triangleright$  These vectors don't change direction when multiplied by A.



## Eigenvalues and Eigenvectors – Properties

- ▶ If  $A\vec{v} = \lambda \vec{v}$ , then  $\vec{v}$  is an eigenvector of A with eigenvalue  $\lambda$ .
- Eigenvectors are defined up to a scalar: if  $\vec{v}$  is an eigenvector, so is  $c\vec{v}$  for any  $c \neq 0$ .
- ► The set of all eigenvectors corresponding to a single eigenvalue forms a vector subspace.
- A square matrix of size  $n \times n$  has at most n eigenvalues (including complex ones and multiplicities).
- ▶ If all eigenvalues of a matrix are positive, the matrix is positive definite.
- ▶ Diagonal matrices (matrixes where for all  $a_{ij} = 0$  where  $i \neg j$ ) have their diagonal elements as eigenvalues, and the standard basis vectors as eigenvectors.
- ► The eigenvalues of a complex Hermitian or real symmetric matrix are real.



## Finding eigenvalues

#### Using a computer:

```
import torch
A = torch.tensor([[2., 0.], [0., 2.]])
eigenvalues, eigenvectors = torch.linalg.eigh(A)
print("Eigenvalues:", eigenvalues)
print("Eigenvectors:\n", eigenvectors)
```

### Revision

In this presentation you learned about:

- ► Vectors.
- ► Matrices.
- ▶ Dot products.
- ▶ Vector-Matrix and Matrix-Matrix multiplication.
- ► Eigen-decomposition.

## Extra: Matrix's determinant

#### Mathematical definition:

ightharpoonup The determinant is a scalar-valued function of the entries of a square matrix A.

The determinant encodes:

- ightharpoonup A is invertibe if  $det(A) \neq 0$
- ightharpoonup Absolute value equals the volume change under the linear transformation of A.

#### Formula:

• for 
$$2 \times 2$$
 case:  $\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$ 

 $\triangleright$  for  $3\times3$  case:

$$\det \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = aei + bfg + cdh - ceg - bdi - ahf.$$

#### Use in AI:

- ► Checking invertibility.
- ► Eigenvalue Problem.



## Extra: Eigenvalue Problem (detailed)

- ► Step 1: Start with  $A\vec{v} = \lambda \vec{v}$ .
- ▶ Step 2: Rewrite as  $(A \lambda I)\vec{v} = 0$ .
- ▶ Step 3: Solve  $det(A \lambda I) = 0$  (this gives a polynomial).
- ▶ Step 4: Find the roots (eigenvalues), then solve the nullspace for each to get eigenvectors.

## Example 6

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \Rightarrow \det \left( \begin{bmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{bmatrix} \right) = (2 - \lambda)^2 - 1 = \lambda^2 - 4\lambda + 3.$$
(22)

- ightharpoonup Roots:  $\lambda = 1, 3$ .
- For  $\lambda = 3$ :  $(A 3I)\vec{v} = 0 \Rightarrow \vec{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ .

For  $\lambda = 1$ :  $(A - I)\vec{v} = 0 \Rightarrow \vec{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .

