

# Analytical Geometry and Linear Algebra. Lecture 3.

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## End of Lecture #2

### Review. Lecture 2

- Part 1. The Dot Product and its properties
  - Norm of a vector
  - Cauchy-Schwarz inequality
  - Triangle Inequality
- Part 2. Vector Cross Product
- Part 3. Matrices (2x2, 3x3).

## Quiz in class

Go to <http://b.socrative.com>

Type Room: **LINAL**

Answer questions.

## Lecture 3. Outline

- Part 1 (recap). Matrices. Transpose, Addition, Scalar multiplication
- Part 2. Matrix multiplication
- Part 3. Determinants. Scalar Triple Product

# Part 1. Matrices

## Definition

Matrix  $A$  is a rectangular table of numbers with  $m$  rows and  $n$  columns.

Example of a  $3 \times 3$  matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

Example of a  $2 \times 3$  matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix}$$

## Different kinds of matrices

$A$  is a  $m \times n$  matrix

- Square ( $m = n$ )
- Rectangular matrix ( $m \neq n$ )
- Symmetric matrix ( $A^T = A$ )
- (Upper) Triangular matrix ( $\forall i, j$ , such that  $i > j : a_{i,j} = 0$ )
- Diagonal matrix ( $\forall i, j$ , such that  $i \neq j : a_{i,j} = 0$ )
- Identity matrix ( $IA = AI = A$ )
- Zero matrix ( $0 + A = A$ )

# Examples

(1) Square matrix : (#rows = #columns)

$$\begin{bmatrix} 3 & -1 & -3 \\ 2 & 4 & 0 \\ -1 & 5 & 6 \end{bmatrix}$$

Main diagonal

(9) Column matrix ( $n$ -vector)

(8) Row matrix ( $n$ -vector):  
[1 7 -3] is a 3-vector

$$\begin{bmatrix} 3 \\ -2 \\ 4 \end{bmatrix} \text{ is a 3-vector}$$

(2) Upper triangular matrix :

$$\begin{bmatrix} 3 & -1 & -3 \\ 0 & 4 & 7 \\ 0 & 0 & 6 \end{bmatrix}$$

(3) Lower triangular matrix :

$$\begin{bmatrix} 3 & 0 & 0 \\ -2 & 4 & 0 \\ 2 & 7 & -6 \end{bmatrix}$$

(5) Diagonal matrix :

$$\begin{bmatrix} 3 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & -6 \end{bmatrix}$$

(6) Identity matrix :

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

(7) Zero matrix :

$$O_{2 \times 2} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, O_{2 \times 3} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Source: <https://medium.com/@nithishraghav/linear-algebra-for-aspiring-data-scientists-part-i-37a9b63c031f>



## Operations. Transpose a matrix

## Transpose of matrix

If  $A$  is an  $m \times n$  matrix, the *transpose*  $A^T$  is an  $n \times m$  matrix defined by  $(A^T)_{ij} = A_{ji}$ .

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

$$\forall A, (A^T)^T = A$$

## Operations. Addition, multiplication by a scalar

Element-wise addition:

$$\begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix} + \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix} = \begin{bmatrix} 1 + a & 4 + d \\ 2 + b & 5 + e \\ 3 + c & 6 + f \end{bmatrix}$$

Properties.  $A, B, C$  are matrices of the same size (!)

- $A + B = B + A$  (commutative)
- $A + (B + C) = (A + B) + C$  (associative)
- $B = \lambda A, \lambda \in \mathbb{R}$  (multiplication by a scalar  $\lambda$ , element-wise)

## Trace of a matrix

Definition of trace of a **square** matrix  $A$

$$Tr(A) = \sum_{i=1}^m a_{ii}$$

$$Tr(A + B) = Tr(A) + Tr(B)$$

$$\forall \lambda \in \mathbb{R}, \quad Tr(\lambda A) = \lambda Tr(A)$$

Linearity of the trace operator means

$$Tr(\alpha A + \beta B) = \alpha Tr(A) + \beta Tr(B)$$

## Part 2. Matrix multiplication

## Matrix multiplication. Definition

Let

$A$  be  $m \times n$  matrix;

$B$  be  $n \times p$  matrix

Then exists  $C = AB$ ,

$C$  must be  $m \times p$  matrix

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj} = \sum_{k=1}^n a_{ik}b_{kj},$$

for  $i = 1, \dots, m$  and  $j = 1, \dots, p$

## Most important!

Before you multiply two matrices  $A$  and  $B$ .  $A$  is  $m \times n$  matrix;  $B$  is  $k \times p$  matrix

- Commit into your memory: **matrix multiplication is not commutative.**

So, in general:

$$AB \neq BA$$

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- **Check sizes** of the two matrices:
  - if you multiply  $AB$  ( $m \times n$ )( $k \times p$ ), then check that  $n = k$

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So, in general:

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- **Check sizes** of the two matrices:
  - if you multiply  $AB$  ( $m \times n$ )( $k \times p$ ), then check that  $n = k$
- **Calculate the size** of the result:
  - if you multiply  $AB$  ( $m \times n$ )( $k \times p$ ), then the result is a  $m \times p$  matrix.



# Illustration

$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ a_7 & a_8 & a_9 \end{bmatrix} \begin{bmatrix} b_1 & b_2 & b_3 \\ b_4 & b_5 & b_6 \\ b_7 & b_8 & b_9 \end{bmatrix} = \begin{bmatrix} c_1 & c_2 & c_3 \\ c_4 & c_5 & c_6 \\ c_7 & c_8 & c_9 \end{bmatrix}$$

# Python Code

A@B

How to calculate the result? Example  $2 \times 2$  matrices

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

$$AB = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} =$$

How to calculate? Example  $2 \times 2$  matrices

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

$$AB = \begin{bmatrix} \mathbf{1} & \mathbf{2} \\ 3 & 4 \end{bmatrix} \begin{bmatrix} \mathbf{a} & b \\ \mathbf{c} & d \end{bmatrix} = \begin{bmatrix} 1a + 2c & * \\ * & * \end{bmatrix}$$

How to calculate the result? Example  $2 \times 2$  matrices

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

$$AB = \begin{bmatrix} \mathbf{1} & \mathbf{2} \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & \mathbf{b} \\ c & \mathbf{d} \end{bmatrix} = \begin{bmatrix} 1a + 2c & 1b + 2d \\ * & * \end{bmatrix}$$

How to calculate the result? Example  $2 \times 2$  matrices

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

$$AB = \begin{bmatrix} 1 & 2 \\ \mathbf{3} & \mathbf{4} \end{bmatrix} \begin{bmatrix} \mathbf{a} & b \\ \mathbf{c} & d \end{bmatrix} = \begin{bmatrix} 1a + 2c & 1b + 2d \\ 3a + 4c & * \end{bmatrix}$$

How to calculate the result? Example  $2 \times 2$  matrices

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

$$AB = \begin{bmatrix} 1 & 2 \\ \mathbf{3} & \mathbf{4} \end{bmatrix} \begin{bmatrix} a & \mathbf{b} \\ c & \mathbf{d} \end{bmatrix} = \begin{bmatrix} 1a + 2c & 1b + 2d \\ 3a + 4c & 3b + 4d \end{bmatrix}$$

Your turn!

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

$$BA = ?$$



Your turn!

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

$$BA = ?$$

True or False?

$$AB = BA?$$

Your turn!

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

$$BA = ?$$

True or False?

$$AB = BA?$$

$$(AB)C = A(BC) = ABC?$$

## Exercise

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 6 \end{bmatrix}, B = \begin{bmatrix} x & u & a \\ y & v & b \\ z & w & c \end{bmatrix}$$

$$AB = ?$$

## Three other ways to think about matrix multiplication

- row-oriented view
- column-oriented view
- layer-oriented view

## Three other ways to think about matrix multiplication

- row-oriented view

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

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 1 \begin{bmatrix} a & b \end{bmatrix} + 2 \begin{bmatrix} c & d \end{bmatrix} \\ 3 \begin{bmatrix} a & b \end{bmatrix} + 4 \begin{bmatrix} c & d \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} 1a & 1b \end{bmatrix} + \begin{bmatrix} 2c & 2d \end{bmatrix} \\ \begin{bmatrix} 3a & 3b \end{bmatrix} + \begin{bmatrix} 4c & 4d \end{bmatrix} \end{bmatrix}$$

Here result is still a  $2 \times 2$  matrix.

It has two rows, but each row is a  $1 \times 2$  vector (!)

## Three other ways to think about matrix multiplication

- column-oriented view

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

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a \begin{bmatrix} 1 \\ 3 \end{bmatrix} + c \begin{bmatrix} 2 \\ 4 \end{bmatrix}, & b \begin{bmatrix} 1 \\ 3 \end{bmatrix} + d \begin{bmatrix} 2 \\ 4 \end{bmatrix} \end{bmatrix} = \left[ \begin{bmatrix} 1a \\ 3a \end{bmatrix} + \begin{bmatrix} 2c \\ 4c \end{bmatrix}, \begin{bmatrix} 1b \\ 3b \end{bmatrix} + \begin{bmatrix} 2d \\ 4d \end{bmatrix} \right]$$

Here result is still a  $2 \times 2$  matrix.

It has two columns, but each column is a  $2 \times 1$  vector (!)

## Three other ways to think about matrix multiplication

- layer-oriented view

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

$$AB = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 1a & 1b \\ 3a & 3b \end{bmatrix} + \begin{bmatrix} 2c & 2d \\ 4c & 4d \end{bmatrix} = \begin{bmatrix} 1a + 2c & 1b + 2d \\ 3a + 4c & 3b + 4d \end{bmatrix}$$

Here result is still a  $2 \times 2$  matrix. It is represented as a sum of 'simpler' matrices.

## Assignment

$$A = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 2 & 1 & 0 & -1 \end{bmatrix}, B = \begin{bmatrix} 0 & 2 \\ 1 & 2 \\ 1 & 2 \\ 1 & 1 \end{bmatrix}$$

$$AB =$$



## Order of operations after transposition

$$(ABCD)^{\top} = D^{\top} C^{\top} B^{\top} A^{\top}$$

Very special and important case

Matrix vector multiplication

$$A\mathbf{x}$$

or

$$\mathbf{x}^T A$$

## Matrix - vector multiplication

Result is always a vector!

$$A\mathbf{x}$$

$$(m \times n)(n \times 1) \rightarrow (m \times 1) \quad \text{is a column-vector}$$

$$\mathbf{x}^T A$$

$$(1 \times m)(m \times n) \rightarrow (1 \times n) \quad \text{is a row-vector}$$

So, we can see that matrix multiplication transforms vectors. Matrix  $A$  is a linear map.

## Matrix as a linear transformation

Again, it is important!  
Matrix  $A$  is a linear map.

Vector  $x$  was a  $(n \times 1)$  column-vector

$$Ax$$

$$(m \times n)(n \times 1) \rightarrow (m \times 1) \text{ column-vector}$$

Result is  $(m \times 1)$  column-vector

$A$  maps vectors in  $\mathbb{R}^n$  to vectors in  $\mathbb{R}^m$

## Examples of transformations. Rotation

## Rotation matrix in 2D

$$A = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

**A rotates any vector  $\mathbf{x} = [x_1, x_2]^\top$  by an angle  $\theta$  counter-clockwise!**

$$A\mathbf{x} = \begin{bmatrix} x_1 \cos(\theta) - x_2 \sin(\theta) \\ x_1 \sin(\theta) + x_2 \cos(\theta) \end{bmatrix}$$

## Code in python

Here we run some code in Colab.

```
https://colab.research.google.com/drive/  
1Kfv4253b5duaP-KjQTk4Ail9rjR45pCX#scrollTo=RNdMstvGEUf0
```

## A very interesting case

What if multiplication  $A\mathbf{w}$  work as follows?

$$A\mathbf{w} = \lambda\mathbf{w}, \quad \lambda \in \mathbb{R}$$

### Example

$$\begin{bmatrix} 2 & 1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 4 \\ 8 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

### Interesting indeed!

$\lambda$  is called eigenvalue

$\mathbf{w}$  is called eigenvector

Break, 5 min.  
Watch some funny video on youtube :)  
<https://youtu.be/BKorP55Aqvg>



## Part 3. Determinants

## Determinant. Concept and application

Notation:  $\det(A)$

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} \text{ is a } 2 \times 2 \text{ determinant,}$$

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} \text{ is a } 3 \times 3 \text{ determinant}$$

- Determinant is a **single** number  $\det(A) \in \mathbb{R}$
- Defined only for square matrices!
- $\det(A) = 0$  if  $A$  contains linearly dependent columns. Matrix in this case is called singular.

## Determinant. Concept and applications

### Applications

- Calculating Area/Volume of shape specified by coordinates in matrix
- Finding matrix inverse (later in this course).

## 2x2 Determinant

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

What if we swap rows of the matrix?

$$\begin{vmatrix} a & a\beta \\ b & b\beta \end{vmatrix} = ?$$

# Examples

$$\begin{vmatrix} 1 & 3 \\ 2 & 1 \end{vmatrix} = ?$$

$$\begin{vmatrix} -1 & 2 \\ 4 & -5 \end{vmatrix} = ?$$

## Examples

$$\begin{vmatrix} 1 & 3 \\ 2 & 1 \end{vmatrix} = ?$$

$$\begin{vmatrix} -1 & 2 \\ 4 & -5 \end{vmatrix} = ?$$

What value of  $\lambda$  makes the following determinant zero?

$$\begin{vmatrix} 1 & 2 \\ 4 & \lambda \end{vmatrix} = ?$$

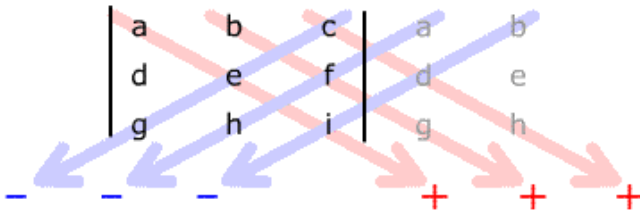
$$\begin{vmatrix} 5-\lambda & -1/3 \\ 3 & 5-\lambda \end{vmatrix} = ?$$

## 3x3 Determinant

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

## 3x3 Determinant

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



$$= aei + bfg + cdh - ceg - afh - bdi$$

Source: <http://thejuniverse.org/PUBLIC/LinearAlgebra/LOLA/detDef/special.html>



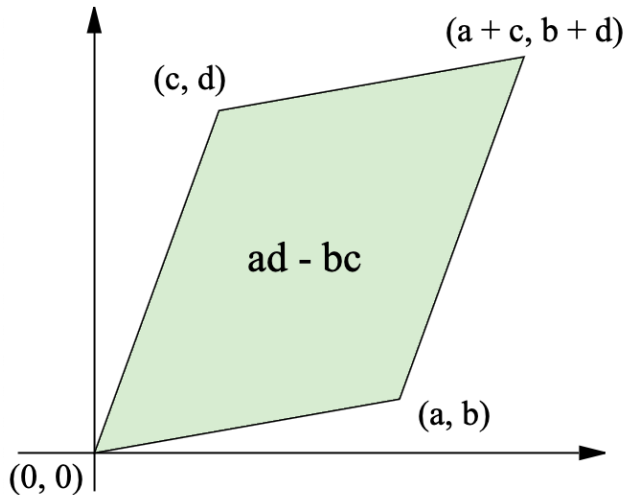
## Examples

$$\begin{vmatrix} 1 & 1 & 2 \\ 2 & 2 & 1 \\ 4 & 3 & 0 \end{vmatrix} = ?$$

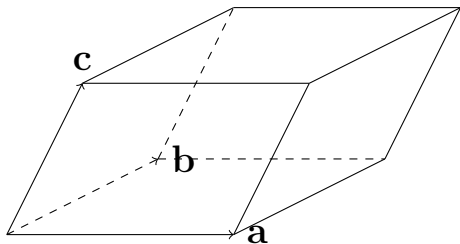
Yes, there exists one single general super formula for calculation of  $\det(A)$  for any arbitrary square matrix  $A$ .

<https://en.wikipedia.org/wiki/Determinant>

# Meaning of the Determinant. Area of a parallelogram

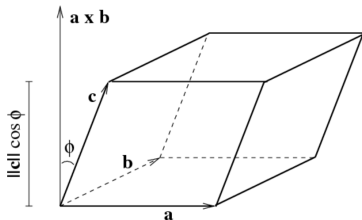


# Meaning of the Determinant. Volume of parallelepiped



$$V = \begin{vmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \end{vmatrix} = \begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix}$$

# Scalar Triple Product



Scalar Triple Product. Definition

$$\mathbf{a} \cdot (\mathbf{b} \times \mathbf{c})$$

$$\mathbf{a} \cdot (\mathbf{b} \times \mathbf{c}) = (\mathbf{a} \times \mathbf{b}) \cdot \mathbf{c}$$

Meaning:  $V = \|\mathbf{a} \times \mathbf{b}\|(\|\mathbf{c}\| \cos(\phi)) = \text{Area of base} * \text{Height}$

## Homework assignment 1: Check the following properties

- $\det(A) = \det(A^\top)$
- $\det(AB) = \det(A)\det(B)$

## Homework assignment 2

Given that  $BC$  and  $CB$  are valid, prove that

$$\text{Tr}(BC) = \text{Tr}(CB)$$

End of Lecture 3

## Useful links

- <https://www.geogebra.org>
- [https://youtu.be/fNk\\_zzaMoSs](https://youtu.be/fNk_zzaMoSs)
- <http://immersivemath.com/ila>