

# Matrices and their operations No. 1

## Multiplication of $2 \times 2$ Matrices

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# Review 1: $2 \times 2$ Matrices

$2 \times 2$  matrices

$$A = (\vec{a}_1 \ \vec{a}_2) = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

A  $2 \times 2$  matrix is given in the following ways.

- (i) Combining two column vectors  $\vec{a}_1, \vec{a}_2 \in \mathbf{R}^2$
- (ii) Combining two row vectors  $\mathbf{a}_1$  and  $\mathbf{a}_2$
- (iii) Giving  $2 \times 2$  components.

**NB**  $a_{ij}$  is used for the component of the  $i$ th row and of the  $j$ th column.

## Review 2

### Multiplication of 2-dim. vectors to $2 \times 2$ matrices

$$\begin{aligned} A \begin{pmatrix} x \\ y \end{pmatrix} &= x\vec{a}_1 + y\vec{a}_2 \\ &= x \begin{pmatrix} a_{11} \\ a_{21} \end{pmatrix} + y \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix} = \begin{pmatrix} xa_{11} + ya_{12} \\ xa_{21} + ya_{22} \end{pmatrix} \\ &= \begin{pmatrix} \mathbf{a}_1 \begin{pmatrix} x \\ y \end{pmatrix} \\ \mathbf{a}_2 \begin{pmatrix} x \\ y \end{pmatrix} \end{pmatrix} \end{aligned}$$

Here we use the multiplication of a row vector and a column vector defined by

$$(\alpha \ \beta) \begin{pmatrix} x \\ y \end{pmatrix} = \alpha x + \beta y$$

## Review 3: Multiplication of two $2 \times 2$ matrices

Take another  $2 \times 2$  matrix

$$B = (\vec{b}_1 \ \vec{b}_2) = \begin{pmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \end{pmatrix}.$$

Then

$$AB = (A\vec{b}_1 \ A\vec{b}_2) = \begin{pmatrix} \mathbf{a}_1 \vec{b}_1 & \mathbf{a}_1 \vec{b}_2 \\ \mathbf{a}_2 \vec{b}_1 & \mathbf{a}_2 \vec{b}_2 \end{pmatrix}$$

# Linear Map defined by using $A$

## A Map defined by $A$

Given a  $2 \times 2$  matrix

$$A = (\vec{a}_1 \ \vec{a}_2) = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix},$$

we can define a map

$$F_A : \mathbf{R}^2 \longrightarrow \mathbf{R}^2$$

$$\begin{pmatrix} s \\ t \end{pmatrix} \mapsto A \begin{pmatrix} s \\ t \end{pmatrix} = s\vec{a}_1 + t\vec{a}_2$$

# Linearity of $F_A$

## Linearity of $F_A$

$F_A$  satisfies the following basic properties called **Linearity**.

- (i)  $F_A(\vec{x} + \vec{y}) = F_A(\vec{x}) + F_A(\vec{y})$
- (ii)  $F_A(\lambda \vec{x}) = \lambda F_A(\vec{x})$
- (iii)  $F_A(\lambda \vec{x} + \mu \vec{y}) = \lambda F_A(\vec{x}) + \mu F_A(\vec{y})$

These three properties are identical to the following.

- (i)  $A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$
- (ii)  $A(\lambda \vec{x}) = \lambda(A\vec{x})$
- (iii)  $A(\lambda \vec{x} + \mu \vec{y}) = \lambda(A\vec{x}) + \mu(A\vec{y})$

Moreover remark that (iii) can be easily derived from (i) and (ii).

In fact

$$\begin{aligned} A(\lambda \vec{x} + \mu \vec{y}) &= A(\lambda \vec{x}) + A(\mu \vec{y}) \\ &= \lambda(A\vec{x}) + \mu(A\vec{y}) \end{aligned}$$

# Proof

## Proof for (i)

$$\begin{aligned} LHS &= A \left( \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right) = A \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \end{pmatrix} \\ &= (x_1 + y_1)\vec{a}_1 + (x_2 + y_2)\vec{a}_2 \\ &= x_1\vec{a}_1 + y_1\vec{a}_1 + x_2\vec{a}_2 + y_2\vec{a}_2 \\ &= (x_1\vec{a}_1 + x_2\vec{a}_2) + (y_1\vec{a}_1 + y_2\vec{a}_2) \\ &= A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + A \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = RHS \end{aligned}$$

## Proof for (ii)

$$\begin{aligned} LHS &= A \begin{pmatrix} \lambda x_1 \\ \lambda x_2 \end{pmatrix} \\ &= (\lambda x_1)\vec{a}_1 + (\lambda x_2)\vec{a}_2 \\ &= \lambda(x_1\vec{a}_1) + \lambda(x_2\vec{a}_2) \\ &= \lambda(x_1\vec{a}_1 + x_2\vec{a}_2) = RHS \end{aligned}$$

# Associativity

**Thanks to** the Linearity, we can prove the following theorem about **Associativity**.

## Theorem: Associativity

Given  $2 \times 2$  matrices  $A$  and  $B$ . Then we have for  $\vec{x} \in \mathbf{R}^2$ .

$$(AB)\vec{x} = A(B\vec{x})$$

In fact,

$$\begin{aligned} RHS &= A(x_1 \vec{b}_1 + x_2 \vec{b}_2) \\ &= x_1(A\vec{b}_1) + x_2(A\vec{b}_2) \\ &= (A\vec{b}_1 \ A\vec{b}_2) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = (AB)\vec{x} = LHS \end{aligned}$$

# Associativity 2

## Theorem: Associativity

Given  $2 \times 2$  matrices  $A$ ,  $B$  and  $C$ . Then

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

# Scalar Multiplication to Matrices

## Scalar Multiplication to $2 \times 2$ Matrices

Given a  $2 \times 2$  matrix

$$A = (\vec{a}_1 \ \vec{a}_2) = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix},$$

we define a scalar multiplication by  $\lambda$  to  $A$  as follows:

$$\lambda A = (\lambda \vec{a}_1 \ \lambda \vec{a}_2) = \begin{pmatrix} \lambda \mathbf{a}_1 \\ \lambda \mathbf{a}_2 \end{pmatrix} = \begin{pmatrix} \lambda a_{11} & \lambda a_{12} \\ \lambda a_{21} & \lambda a_{22} \end{pmatrix}$$

## Theorem

- **(i)**  $(\lambda A)\vec{x} = \lambda(A\vec{x}) = A(\lambda\vec{x})$
- **(ii)**  $(\lambda A)B = \lambda(AB) = A(\lambda B)$

The proof for **(i)** is given as follows:

$$\begin{aligned}(\lambda A)\vec{x} &= (\lambda \vec{a}_1 \ \lambda \vec{a}_2)\vec{x} \\&= (\lambda x_1)\vec{a}_1 + (\lambda x_2)\vec{a}_2 \\&= \lambda(x_1 \vec{a}_1) + \lambda(x_2 \vec{a}_2) \\&= \lambda(x_1 \vec{a}_1 + x_2 \vec{a}_2) = \lambda(A\vec{x})\end{aligned}$$

Moreover the property **(ii)** is derived easily from **(i)**.

## Theorem

- (iii)  $(\lambda + \mu)A = \lambda A + \mu A$
- (iv)  $(\lambda\mu)A = \lambda(\mu A)$
- (v)  $1A = A$  and  $0A = O_2$

These properties can be derived from the following corresponding properties for vectors. It is necessary to define the addition of matrices to understand (iii), and we will put it off for a couple of weeks.

- (iii)  $(\lambda + \mu)\vec{a} = \lambda\vec{a} + \mu\vec{a}$
- (iv)  $(\lambda\mu)\vec{a} = \lambda(\mu\vec{a})$
- (v)  $1\vec{a} = \vec{a}$  and  $0\vec{a} = \vec{0}$

# Special Matrices

## $O_2$ Zero Matrix

$$O_2 = (\vec{0} \ \vec{0}) = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

is called the zero matrix. It satisfies the identity

$$O_2 X = X O_2 = O_2$$

$O_2 X = O_2$  follows from

$$O_2 \vec{x} = (\vec{0} \ \vec{0}) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x_1 \vec{0} + x_2 \vec{0} = \vec{0}$$

and  $X O_2 = O_2$  from

$$X \vec{0} = (\vec{x}_1 \ \vec{x}_2) \begin{pmatrix} 0 \\ 0 \end{pmatrix} = 0 \vec{x}_1 + 0 \vec{x}_2 = \vec{0}$$

# Special Matrices 2

## $I_2$ Identity Matrix

$$I_2 = (\vec{e}_1 \ \vec{e}_2) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

is called the **Identity Matrix**. It enjoys for a  $2 \times 2$  matrix  $X$ , the identity

$$XI_2 = I_2X = X$$

It follows from the identities

$$(\vec{a} \ \vec{b})\vec{e}_1 = 1\vec{a} + 0\vec{b} = \vec{a}, \quad (\vec{a} \ \vec{b})\vec{e}_2 = 0\vec{a} + 1\vec{b} = \vec{b}$$

that

$$XI_2 = (\vec{x}_1 \ \vec{x}_2)(\vec{e}_1 \ \vec{e}_2) = (\vec{x}_1 \ \vec{x}_2) = X$$

Moreover  $I_2X = X$  from

$$(\vec{e}_1 \ \vec{e}_2) \begin{pmatrix} x \\ y \end{pmatrix} = x\vec{e}_1 + y\vec{e}_2 = x \begin{pmatrix} 1 \\ 0 \end{pmatrix} + y \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix}$$

# Inverse matrix

We have the identity

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} ad - bc & 0 \\ 0 & ad - bc \end{pmatrix}$$

## Cofactor Matrix

For  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , its cofactor matrix is defined by

$$\tilde{A} = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Then we have the identity

$$A\tilde{A} = \tilde{A}A = |A| \cdot I_2$$

# Inverse Matrix 2

Assume that  $|A| \neq 0$ . Then multiply by  $\frac{1}{|A|}$  and get

$$A \cdot \frac{1}{|A|} \tilde{A} = \frac{1}{|A|} \tilde{A} \cdot A = I_2$$

## Inverse Matrix

In case  $|A| \neq 0$ , the **Inverse Matrix** of  $A$  is defined by

$$A^{-1} = \frac{1}{|A|} \tilde{A} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

# Regularity of Matrices, Uniqueness of Inverse

## Regularity of Matrices

A  $2 \times 2$  matrix  $A$  is called **regular** if there exists another  $2 \times 2$  matrix  $X$  satisfying

$$AX = XA = I_2$$

In this situation  $X$  is called the **inverse** of  $A$ .

- (i) If  $|A| \neq 0$ ,  $A$  is regular.
- (ii) (**Uniqueness of the inverse**) Assume

$$AX = XA = I_2, \quad AY = YA = I_2$$

Then  $X = Y$ . In fact, from  $AX = I_2$  multiplied by  $Y$  from the left follows

$$Y(AX) = YI_2 = Y$$

On the other hand,  $Y(AX) = (YA)X = I_2X = X$ .  
Accordingly  $X = Y$ .

In case  $|A| = 0$

### Theorem A

Let  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  a  $2 \times 2$  matrix with  $|A| = 0$ . Then there exists  $\vec{v} \neq \vec{0}$  satisfying  $A\vec{v} = \vec{0}$ .

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} d \\ -c \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} -b \\ a \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

- **(i)** In case  $d \neq 0$  or  $c \neq 0$ ,  $\begin{pmatrix} d \\ -c \end{pmatrix} \neq \vec{0}$  AND  $A \begin{pmatrix} d \\ -c \end{pmatrix} = \vec{0}$ .
- **(ii)** In case  $b \neq 0$  or  $c \neq 0$ ,  $\begin{pmatrix} -b \\ a \end{pmatrix} \neq \vec{0}$  AND  $A \begin{pmatrix} -b \\ a \end{pmatrix} = \vec{0}$ .
- **(iii)** Not (i) AND Not (ii). Then  $A = O_2$ .

# Equivalent conditions for regularity

If  $A$  is regular, then

$$A\vec{v} = \vec{0} \text{ implies } \vec{v} = \vec{0}.$$

In fact by multiplying  $A^{-1}$  to  $A\vec{v} = \vec{0}$  to get

$$A^{-1}A\vec{v} = A^{-1}\vec{0} \text{ namely } \vec{v} = I_2\vec{v} = \vec{0}$$

Thus it follows from Theorem A that if  $|A| = 0$  then  $A$  is not regular.

## Theorem B

The following (i), (ii) and (iii) are equivalent for a  $2 \times 2$  matrix  $A$ .

- (i)  $A$  is regular.
- (ii)  $A\vec{v} = \vec{0} \Rightarrow \vec{v} = \vec{0}$
- (iii)  $|A| \neq 0$ .

# Proof for Theorem B

- $(i) \Rightarrow (iii)$  The contraposition  $\text{Not } (iii) \Rightarrow \text{Not } (i)$  is already shown.
- $(i) \Rightarrow (ii)$  Already shown.
- $(iii) \Rightarrow (i)$  Already shown.
- The contraposition  $\text{Not } (iii) \Rightarrow \text{Not } (ii)$  is given in Theorem A.

# Addition of two $2 \times 2$ Matrices

## Definition

Given two  $2 \times 2$  matrices

$$A = (\vec{a}_1 \ \vec{a}_2) = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

and

$$B = (\vec{b}_1 \ \vec{b}_2) = \begin{pmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \end{pmatrix} = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$$

Then the  $2 \times 2$  matrix  $A + B$  is defined by

$$A + B = (\vec{a}_1 + \vec{b}_1 \ \vec{a}_2 + \vec{b}_2) = \begin{pmatrix} \mathbf{a}_1 + \mathbf{b}_1 \\ \mathbf{a}_2 + \mathbf{b}_2 \end{pmatrix} = \begin{pmatrix} a_{11} + b_{11} & a_{12} + b_{12} \\ a_{21} + b_{21} & a_{22} + b_{22} \end{pmatrix}$$

# Basic Properties (1)

## Basic Properties (1)

- (i)  $(A + B) + C = A + (B + C)$
- (ii)  $A + O_2 = O_2 + A = A$
- (iii)  $A + B = B + A$
- (iv)  $\lambda(A + B) = \lambda A + \lambda B$
- (v)  $(\lambda + \mu)A = \lambda A + \mu A$

(i) follows from  $(\vec{a} + \vec{b}) + \vec{c} = \vec{a} + (\vec{b} + \vec{c})$ .

(ii) follows from  $\vec{a} + \vec{0} = \vec{0} + \vec{a} = \vec{a}$ .

(iii) follows from  $\vec{a} + \vec{b} = \vec{b} + \vec{a}$ .

(v) follows from  $(\lambda + \mu)\vec{a} = \lambda\vec{a} + \mu\vec{a}$ .

(iv) is proved as follows.

$$\begin{aligned} LHS &= \lambda(\vec{a}_1 + \vec{b}_1) \vec{a}_2 + \vec{b}_2 = (\lambda(\vec{a}_1 + \vec{b}_1) \lambda(\vec{a}_2 + \vec{b}_2)) \\ &= (\lambda\vec{a}_1 + \lambda\vec{b}_1) \lambda\vec{a}_2 + \lambda\vec{b}_2 = (\lambda\vec{a}_1 \lambda\vec{a}_2) + (\lambda\vec{b}_1 \lambda\vec{b}_2) = RHS \end{aligned}$$

## Basic Properties (2)

### Basic Properties (2)

- (vi)  $A(B + C) = AB + AC$
- (vii)  $(B + C)A = BA + CA$

(vi) follows from  $A(\vec{b} + \vec{c}) = A\vec{b} + A\vec{c}$ . In fact

$$\begin{aligned} LHS &= A(\vec{b}_1 + \vec{c}_1 \ \vec{b}_2 + \vec{c}_2) = (A(\vec{b}_1 + \vec{c}_1) \ A(\vec{b}_2 + \vec{c}_2)) \\ &= (A\vec{b}_1 + A\vec{c}_1 \ A\vec{b}_2 + A\vec{c}_2) = (A\vec{b}_1 \ A\vec{b}_2) + (A\vec{c}_1 \ A\vec{c}_2) = RHS \end{aligned}$$

(vii) follows from the identity  $(B + C)\vec{a} = B\vec{a} + C\vec{a}$  which is derived as follows.

$$\begin{aligned} LHS &= (\vec{b}_1 + \vec{c}_1 \ \vec{b}_2 + \vec{c}_2) \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} = a_1(\vec{b}_1 + \vec{c}_1) + a_2(\vec{b}_2 + \vec{c}_2) \\ &= \dots = (a_1\vec{b}_1 + a_2\vec{b}_2) + (a_1\vec{c}_1 + a_2\vec{c}_2) = RHS \end{aligned}$$