

# 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 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:

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

# Scalar Multiplication to Matrices

## 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)**.

# Other Basic Properties of Scalar Multiplication

## 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 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  $0\vec{a} = \vec{0}$

## $O_2$ 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

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

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

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

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)**  $A\vec{v} = \vec{0} \quad \Rightarrow \quad \vec{v} = \vec{0}$
- **(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.

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

# 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

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

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

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