

# 3.1 Operations on Matrices

**Quote.** "By natural selection our mind has adapted itself to the conditions of the external world. It has adopted the geometry most advantageous to the species or, in other words, the most convenient"  
Jules Henri Poincaré (1854-1912)

## Vocabulary.

- Matrix - a rectangular array of numbers.



## 1. Matrices

A **matrix** is a rectangular array of numbers (called **entries**):

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} = A$$

The size of this matrix is

$M \times N$  (don't backwards them up)

$M = \# \text{ rows}$   
 $N = \# \text{ columns}$

To specify the names of entries and the size of the matrix we will sometimes use the notation:

$$A = [a_{ij}]_{m \times n} = [a_{ij}]$$

or symbol  $(A)_{ij}$  represents the entry of  $A$  in row  $i$  and column  $j$ . Two matrices are equal if they are the same size and all of the entries are the same.

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

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

(all #s have to be in the same place to be equal)

$$(A)_{21} = 3$$

(2nd row, 1st col)

$$A_{22} = 4$$

A **square matrix** is a matrix with the same number of rows and columns:

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \quad n=m$$

The entries  $a_{11}, a_{22}, \dots, a_{nn}$  are called the **diagonal entries** or the **main diagonal**.

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = A_{3 \times 3}$$

(diagonal entries are 1, 5, 9)

## 2. Diagonal Matrices

A square matrix in which all entries off the main diagonal are zero is called a **diagonal matrix**.

Example:  $D = \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}$

## 3. Triangular matrices

- A square matrix in which all entries above the main diagonal are zero is called **lower triangular** (LT), and a square matrix in which all the entries below the main diagonal are zero is called **upper triangular** (UT).
- A matrix that is either upper triangular or lower triangular is called **triangular** (T).

Example: Which one of the following matrices is upper/lower triangular?

(Gauss Elim)  
upper triangular

→  $\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ 0 & a_{22} & a_{23} & a_{24} \\ 0 & 0 & a_{33} & a_{34} \\ 0 & 0 & 0 & a_{44} \end{bmatrix}$

$\begin{bmatrix} a_{11} & 0 & 0 & 0 \\ a_{21} & a_{22} & 0 & 0 \\ a_{31} & a_{32} & a_{33} & 0 \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$  lower triangular

Example: How can we express these properties algebraically?

- $A$  is UT if  $a_{ij} = 0$ , for every  $i > j$ .
- $A$  is LT if  $a_{ij} = 0$ , for every  $j > i$ .

Example: Can a matrix be both lower and upper triangular?

Yes, diagonal matrices are both.

$A^{-1}$

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$

$I_{3 \times 3} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$A \cdot I = A$

$I \cdot A = A$

$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$I = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

Reg World

$6 \cdot 1 = 6$  (Multiplicative identity element)

$1 \cdot 6 = 6$  (multiplicative)

$6 \cdot \frac{1}{6} = 1$  (inverse)

$6 \cdot (-1)$  (does not mean  $\frac{1}{x}$ )

(same thing to matrix)

(not reciprocal)

4. **Equality of matrices:** Two matrices  $A$  and  $B$  are equal if they have the same size and their corresponding entries are equal, that is, if  $A = [a_{ij}]_{m \times n}$  and  $B = [b_{ij}]_{r \times s}$  then

$$A = B \iff \begin{matrix} m=r, n=s \\ \text{if and only if} \end{matrix}$$

$$a_{ij} = b_{ij} \quad \forall i, m, n \text{ for all}$$

5. **Addition, subtraction and scalar multiplication of matrices.** Addition, subtraction and scalar multiplication:

If matrices  $A$  and  $B$  have the same size and  $c$  is a scalar then

$$(A+B)_{ij} = (A)_{ij} + (B)_{ij} \quad (A-B)_{ij} = (A)_{ij} - (B)_{ij} \quad (cA)_{ij} = c(A)_{ij}$$

If they **do not** have the same size  $A+B$  and  $A-B$  are *undefined*.

*Example.* Given matrices

$$A = \begin{bmatrix} 2 & 0 \\ 7 & -5 \\ 4 & 3 \end{bmatrix}, B = \begin{bmatrix} 4 & -3 & 0 \\ 1 & 5 & -1 \end{bmatrix}, C = \begin{bmatrix} 1 & 2 & 3 \\ 0 & -1 & 0 \end{bmatrix}, D = \begin{bmatrix} 6 & 3 \\ 4 & 5 \\ 9 & 8 \end{bmatrix}$$

Evaluate  $A+B$ ,  $A+D$ ,  $B-D$ ,  $C-B$ ,  $5A$  and  $(-2)B$

$$A+B = \text{undefined}$$

$$A+D = \begin{bmatrix} 2 & 0 \\ 7 & -5 \\ 4 & 3 \end{bmatrix} + \begin{bmatrix} 6 & 3 \\ 4 & 5 \\ 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 3 \\ 11 & 0 \\ 13 & 11 \end{bmatrix}$$

$$B-D = \text{undefined}$$

$$(-B) = \begin{bmatrix} -4 & 3 & 0 \\ -1 & -5 & 1 \end{bmatrix}$$

$$5A = 5 \begin{bmatrix} 2 & 0 \\ 7 & -5 \\ 4 & 3 \end{bmatrix} = \begin{bmatrix} 10 & 0 \\ 35 & -25 \\ 20 & 15 \end{bmatrix}$$

## 6. List of row and column vectors

Any matrix can be thought of as

- (a) a list (row) of columns vectors; or
- (b) a list (column) of row vectors.

In general, if  $A = [a_{ij}]_{m \times n}$ , we can write

$$A = [\mathbf{c}_1(A) \quad \mathbf{c}_2(A) \quad \dots \quad \mathbf{c}_n(A)] = \begin{bmatrix} \mathbf{r}_1(A) \\ \mathbf{r}_2(A) \\ \vdots \\ \mathbf{r}_m(A) \end{bmatrix}$$

where  $\mathbf{r}_i(A)$  is the  $i$ -th row of  $A$  and  $\mathbf{c}_j(A)$  is the  $j$ -th column of  $A$ .

Example. What are the column and row vectors of  $A = \begin{bmatrix} 3 & 0 & 7 & 5 \\ 2 & 9 & 6 & 0 \\ 1 & 8 & 1 & 4 \end{bmatrix}$

$$A = [\vec{c}_1 \mid \vec{c}_2 \mid \vec{c}_3 \mid \vec{c}_4] \quad \vec{c}_1 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \quad \vec{c}_2 = \begin{bmatrix} 0 \\ 9 \\ 8 \end{bmatrix}$$

$$A = \begin{bmatrix} \vec{r}_1 \\ \vec{r}_2 \\ \vec{r}_3 \end{bmatrix} \quad \vec{r}_1 = [3 \ 0 \ 7 \ 5] \quad \vec{r}_2 = [2 \ 9 \ 6 \ 0] \quad \vec{r}_3 = [1 \ 8 \ 1 \ 4]$$

$\Rightarrow A = \begin{bmatrix} 3 & 0 & 7 & 5 \\ 2 & 9 & 6 & 0 \\ 1 & 8 & 1 & 4 \end{bmatrix}$

## 7. Span and linear independence for Matrices

These follow directly from the same definitions we had for vectors.

Let  $\mathcal{B} = \{A_1, \dots, A_k\}$  be a set of matrices.

The **span** of  $\mathcal{B}$  is defined as  $\text{Span } \mathcal{B} = \{t_1 A_1 + \dots + t_k A_k\}$

The set  $\mathcal{B}$  is linearly independent if the only solution to  $c_1 A_1 + \dots + c_k A_k = 0$  is  $c_1 = \dots = c_k = 0$ .

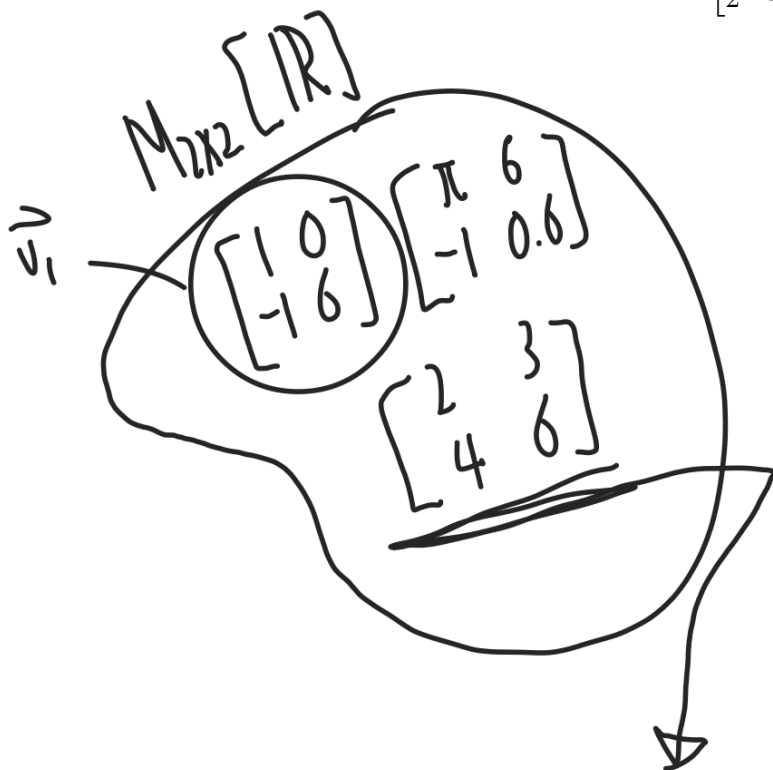
## 8. Example

$$\text{Let } \mathcal{B} = \left\{ \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 3 & -1 \end{bmatrix} \right\}$$

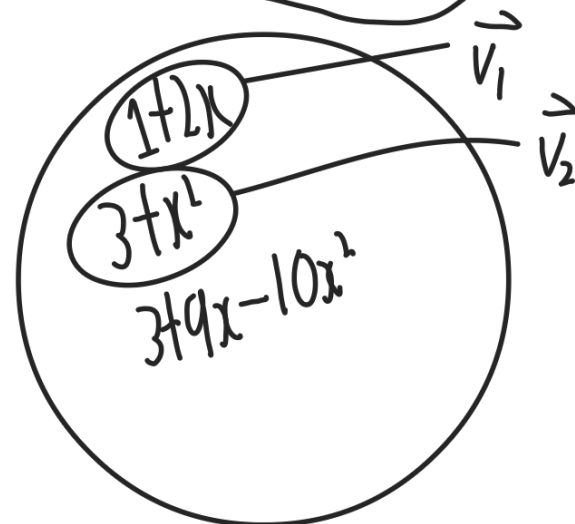
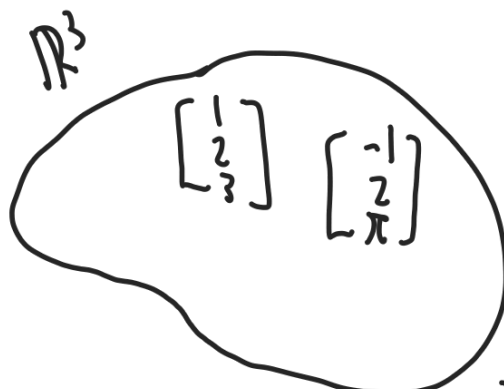
and determine if the set is linearly independent. Is  $\begin{bmatrix} 2 & 3 \\ 2 & -3 \end{bmatrix}$  in the Span  $\mathcal{B}$ ?

In the textbook

Vector Space



Basis  $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$



## 9. Transpose

If  $A$  is an  $m \times n$  matrix, then the **transpose** of  $A$ , denoted as  $A^T$ , is the  $n \times m$  matrix created by making the rows of  $A$  into columns. We can write this as  $(A^T)_{ij} = (A)_{ji}$ .

Example. Suppose  $A = \begin{bmatrix} -3 & 2 & 1 \\ 0 & 2 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} 8 & 7 & 1 \end{bmatrix}$ . Find  $A^T$ ,  $B^T$  and  $(B^T)^T$ .

$$A^T = \begin{bmatrix} -3 & 0 \\ 2 & 2 \\ 1 & 4 \end{bmatrix} \quad B^T = \begin{bmatrix} 8 \\ 7 \\ 1 \end{bmatrix} \quad (B^T)^T = B$$

$\gg A$   
 $\gg A^T$

## 10. Matrix vector multiplication *(Cannot stress how important this is)*

Example. Consider the following linear system

$$\begin{aligned} 3x_1 - 7x_2 + x_3 &= 4 \\ 5x_1 + x_2 + 2x_3 &= 2 \end{aligned}$$

We can naturally associate three matrices with the system:

$$A = \begin{bmatrix} 3 & -7 & 1 \\ 5 & 1 & 2 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, \quad \text{and } \mathbf{b} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$

$A$  is usually called the **coefficient matrix** of the system.

We want to define the product  $A\mathbf{x}$  in such a way that the  $m = 2$  linear equations can be rewritten in a single matrix equation  $A\mathbf{x} = \mathbf{b}$ .

$$A\vec{x} = \vec{b}$$

Note that  $m = 2$  linear equations can be written as a single matrix equation as follows:

$$\begin{bmatrix} 3 & -7 & 1 \\ 5 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$

$$A \cdot \vec{x} = \vec{b}$$

*(Dot Product)*

$$3x_1 - 7x_2 + x_3 = 4$$

$$5x_1 - x_2 + 2x_3 = 2$$

$$\left[ \begin{array}{ccc|c} 3 & -7 & 1 & 4 \\ 5 & 1 & 2 & 2 \end{array} \right]$$

The product  $A\mathbf{x}$  (definition):

Suppose  $A$  is an  $m \times n$  matrix and  $\mathbf{x}$  is an  $n \times 1$  column vector. The **product**  $A\mathbf{x}$  is defined as:

$$A\mathbf{x} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{bmatrix}$$

**Remark:** The result of multiplication is  $m \times 1$  column vector. The product is only defined when the number of columns of  $A$  is the same as the number of rows of  $\mathbf{x}$ .

Example:

$$\begin{bmatrix} 1 & 2 & -1 \\ 0 & 3 & 2 \\ 4 & 5 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 \\ 3 \\ 13 \end{bmatrix}$$

Handwritten notes:  $(1)(2) + (2)(1) + (-1)(0)$ ,  $(0)(2) + (3)(1) + (2)(0)$ ,  $(4)(2) + (5)(1) + (3)(0)$ . Red arrows point to the dimensions:  $3 \times 3$  (3 row, 3 col),  $3 \times 1$ , and  $3 \times 1$ . A red 'X' is over the second vector. A red note says "(Don't get it backwards)".

We can also view the product  $A\mathbf{x}$  as a linear combination of column vectors of  $A$ :

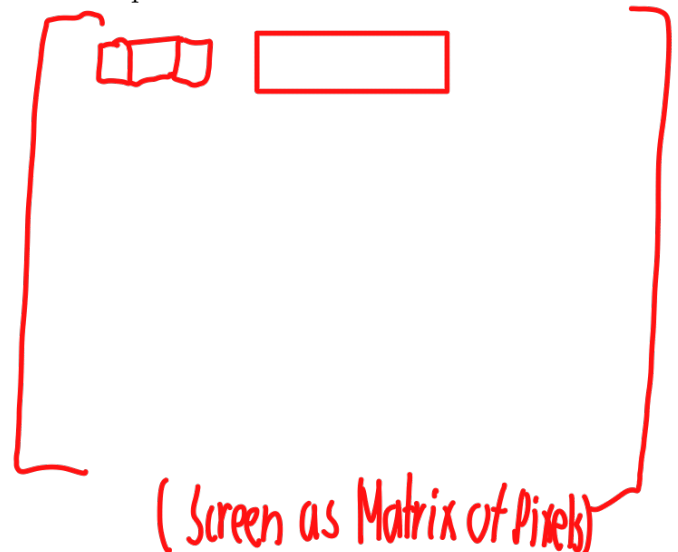
$$A\mathbf{x} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{bmatrix} = x_1 \mathbf{c}_1(A) + x_2 \mathbf{c}_2(A) + \dots + x_n \mathbf{c}_n(A)$$

Ex.  $\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 4 \\ 7 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 5 \\ 8 \end{bmatrix} + x_3 \begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix}$

Handwritten notes:  $x_1 + 2x_2 + 3x_3$ . Red arrows show the expansion of the matrix product into a sum of column vectors.

Or we can view the product  $A\mathbf{x}$  as a column vector composed of dot products of row vectors of  $A$  and  $\mathbf{x}$ :

$$A\mathbf{x} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1(A) \cdot \mathbf{x} \\ \mathbf{r}_2(A) \cdot \mathbf{x} \\ \vdots \\ \mathbf{r}_m(A) \cdot \mathbf{x} \end{bmatrix}$$



Example. Evaluate the following using all three methods:  $(5)(3) + (0)(-2) + (2)(10) = 35$

$$\begin{bmatrix} 5 & 0 & 2 \\ -1 & 4 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \\ 10 \end{bmatrix} = \begin{bmatrix} 35 \\ -1 \end{bmatrix}$$

$$(2 \times 3)(3 \times 1)$$

Mid #s match =  $(2 \times 1)$

## 11. The product $AB$

Suppose  $A$  is an  $m \times n$  matrix and  $B$  is an  $n \times p$  matrix. The **product**  $AB$  is the  $m \times p$  matrix defined as

$$[Ac_1(B) \quad Ac_2(B) \quad \dots \quad Ac_p(B)]$$

Example. Given  $A = \begin{bmatrix} 4 & 0 & 5 \\ 3 & -2 & 8 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 10 & -2 & 0 \\ 0 & 6 & 1 & 5 \\ 2 & 2 & -7 & 9 \end{bmatrix}$ , find  $AB$  and  $BA$ .

$$AB = \begin{bmatrix} 4 & 0 & 5 \\ 3 & -2 & 8 \end{bmatrix} \begin{bmatrix} 1 & 10 & -2 & 0 \\ 0 & 6 & 1 & 5 \\ 2 & 2 & -7 & 9 \end{bmatrix} =$$

$$(2 \times 3) (3 \times 4)$$

$$2 \times 4$$

$$= \begin{bmatrix} 14 & 50 & -43 & 45 \\ 19 & 34 & -64 & 62 \end{bmatrix}$$

$$(3)(-2) + (-2)(1) + (8)(-7)$$

$$BA = \begin{bmatrix} 1 & 10 & -2 & 0 \\ 0 & 6 & 1 & 5 \\ 2 & 2 & -7 & 9 \end{bmatrix} \begin{bmatrix} 4 \\ 0 \\ 5 \\ 3 \end{bmatrix}$$

No!

$$2 \times 3$$

$AB \neq BA$   
Not commutative



$$2 \cdot 3 = 3 \cdot 2$$

$$AB \neq BA$$

$$\cancel{X} \cdot AB = A \cancel{X}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^A \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}^B = \begin{bmatrix} 1 & 5 \\ 3 & 9 \end{bmatrix}^{AB}$$

$$\begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}^B \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^A = \begin{bmatrix} -2 & -2 \\ 9 & 12 \end{bmatrix}^{BA}$$

12. Linearity properties If  $A$  is an  $m \times n$  matrix,  $\mathbf{u}$  and  $\mathbf{v}$  are vectors in  $\mathbb{R}^n$ , and  $c$  is a scalar, then

(a)  $A(c\mathbf{u}) = c(A\mathbf{u})$

(b)  $A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$

$\longleftrightarrow A(c\vec{u}) = (A\vec{u})$

$A(\vec{u} + \vec{v}) = A\vec{u} + A\vec{v}$

• Matrices are...

13. **IMPORTANT** properties of matrix multiplication

- Matrix multiplication is NOT commutative.
- The cancellation law is NOT valid.
- $AB = 0$  does not mean that one of  $A$  or  $B$  is zero.

$AB \neq BA$

$AB = AC$

Not valid!

$AB = 0$  does not mean  $A = 0$  or  $B = 0$

$xy = 0$

It does NOT mean that  $B = C$

14. The Identity Matrix

$I_{2 \times 1} = I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

plays the role of 1.

$A \cdot I = A$

• exception

$I \cdot A = A$

• multiplicative identity element