

---

## Definitions and key facts for section 2.1

---

Recall a **matrix**  $A$  is an  $m \times n$  rectangular array of numbers with  $m$  rows and  $n$  columns. We label the entry of the  $i$ th row and  $j$ th column  $a_{ij}$  and label the  $j$ th column  $\mathbf{a}_j$  so that

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1j} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2j} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots & & \vdots \\ a_{i1} & a_{i2} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \vdots & & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mj} & \cdots & a_{mn} \end{bmatrix} = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_j \quad \cdots \quad \mathbf{a}_n] = [a_{ij}] \quad \text{where } \mathbf{a}_j = \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{ij} \\ \vdots \\ a_{mj} \end{bmatrix}.$$

The **diagonal entries** of  $A$  are  $a_{11}, a_{22}, \dots$  and they form the **main diagonal** of  $A$ . A matrix whose non-diagonal entries are zero is called a **diagonal matrix**.

A matrix with all zero entries is a **zero matrix** and is usually denoted by  $0$ .

The **transpose** of an  $m \times n$  matrix  $A$  is the  $n \times m$  matrix  $A^T$  whose columns are the formed from the corresponding rows of  $A$ .

---

### The algebra of matrices:

- two  $m \times n$  matrices  $A$  and  $B$  are **equal**, written  $A = B$ , if their entries are equal;
- the **sum** two  $m \times n$  matrices  $A$  and  $B$  is the matrix whose entries are the sum of the entries of  $A$  and  $B$ , written  $A + B$ ;
- and the **scalar multiple** an  $m \times n$  matrix  $A$  by a scalar  $c$  (a real number) is the matrix  $c\mathbf{A}$  obtaining by multiplying each entry of  $A$  by  $c$ .

**Basic algebraic properties:** Let  $A, B$  and  $C$  be matrices of the same size and  $c, d$  be scalars.

- |                                |                         |
|--------------------------------|-------------------------|
| 1. $A + B = B + A$             | 4. $c(A + B) = cA + cB$ |
| 2. $(A + B) + C = A + (B + C)$ | 5. $(c + d)A = cA + dA$ |
| 3. $A + 0 = A$                 | 6. $c(dA) = (cd)A$      |
- 

### Matrix multiplication

Given a  $m \times n$  matrix  $A$  and a  $n \times p$  matrix  $B$ ,

$$A = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n] \quad B = [\mathbf{b}_1 \quad \mathbf{b}_2 \quad \cdots \quad \mathbf{b}_p]$$

we define the **product** of  $A$  and  $B$  to be the following  $m \times p$  matrix

$$AB = A[\mathbf{b}_1 \quad \mathbf{b}_2 \quad \cdots \quad \mathbf{b}_p] = [A\mathbf{b}_1 \quad A\mathbf{b}_2 \quad \cdots \quad A\mathbf{b}_p]$$

**Composition of linear transformations** If  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $S : \mathbb{R}^p \rightarrow \mathbb{R}^n$  are linear transformations with standard matrices  $A$  and  $B$  respectively, then

$$T(S(\mathbf{x})) = AB\mathbf{x} \text{ for all } \mathbf{x} \text{ in } \mathbb{R}^p.$$

### Computing $AB$

We compute the product  $AB$  by either of two methods:

- Using the definition, we compute each column of  $AB$  as a linear combination of the columns of  $A$  using weights from the corresponding column of  $B$ .
- Using the **row-column rule**: if  $(AB)_{ij}$  is the entry of  $AB$  in the  $i$ th row and  $j$ th column then

$$(AB)_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj}.$$

**Properties of matrix multiplication:** Let  $A$  be an  $m \times n$  matrix, and let  $B$  and  $C$  have sizes such that the each following expression is defined.

1.  $A(BC) = (AB)C$
2.  $A(B + C) = AB + AC$
3.  $(B + C)A = BA + CA$
4.  $r(AB) = (rA)B = A(rB)$   
for any scalar  $r$
5.  $I_m A = A = A I_n$

### Warnings for matrix multiplication:

In general, a few familiar properties of multiplication **do not** hold for matrices:

1. In general,  $AB \neq BA$ .
2. In general, if  $AB = AC$ , that does not guarantee  $B = C$ .
3. If  $AB = 0$  we cannot conclude in general that either  $A$  or  $B$  is 0

---

If  $A$  is an  $n \times n$  matrix, we define  $A^k$  to be the product of  $k$ -many copies of  $A$

$$A^k = \underbrace{AA \cdots A}_k \text{ and } A^0 = I_n.$$

### Properties of the transpose

With  $A$  and  $B$  of appropriate sizes, we have the following.

1.  $(A^T)^T = A$
2.  $(A + B)^T = A^T + B^T$
3.  $(cA)^T = cA^T$  for any scalar  $c$ .
4.  $(AB)^T = B^T A^T$ .