

## Matrix multiplication

Recall our definition of the product  $\mathbf{AB}$  of two matrices.

**Definition.** If  $\mathbf{A}$  is an  $m \times n$  matrix and  $\mathbf{B}$  is an  $n \times p$  matrix, then the product  $\mathbf{AB}$  is the matrix

$$\mathbf{AB} = \left[ \begin{array}{c|c|c|c} \mathbf{AB}_1 & \mathbf{AB}_2 & \dots & \mathbf{AB}_p \end{array} \right],$$

where  $\mathbf{B}_j$  represents the  $j$ th column of  $\mathbf{B}$ .

**Row-column dot product definition:** The columns of  $\mathbf{AB}$  are linear combinations of the columns of  $\mathbf{A}$ . In fact, consider the  $j$ th column of  $\mathbf{AB}$ .

**Row-column rule:**  $(\mathbf{AB})_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj}$

**Definition.** The  $n \times n$  (square) matrix

$$\mathbf{I}_n = \begin{bmatrix} 1 & 0 & \dots & \dots & 0 \\ 0 & 1 & \dots & \dots & 0 \\ \vdots & \vdots & & & \vdots \\ \vdots & \vdots & & & \vdots \\ 0 & 0 & \dots & \dots & 1 \end{bmatrix}$$

with 1's down the diagonal and 0's everywhere else is called the  $n \times n$  *identity matrix*.

**Theorem 2.** Let  $\mathbf{A}$  be an  $m \times n$  matrix, and let  $\mathbf{B}$  and  $\mathbf{C}$  be matrices of appropriate sizes. Then

1.  $\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$
2.  $\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$
3.  $(\mathbf{B} + \mathbf{C})\mathbf{A} = \mathbf{BA} + \mathbf{CA}$
4.  $r(\mathbf{AB}) = (r\mathbf{A})\mathbf{B} = \mathbf{A}(r\mathbf{B})$  for any scalar  $r$
5.  $\mathbf{I}_m\mathbf{A} = \mathbf{A} = \mathbf{A}\mathbf{I}_n$

**Three warnings.**

1.  $\mathbf{AB}$  does not always equal  $\mathbf{BA}$ .
2.  $\mathbf{AB} = \mathbf{AC}$  does not necessarily imply that  $\mathbf{B} = \mathbf{C}$ .
3.  $\mathbf{AB} = \mathbf{0}$  does not necessarily imply that  $\mathbf{A} = \mathbf{0}$  or  $\mathbf{B} = \mathbf{0}$ .

We will occasionally need to use the transpose of a matrix.

**Definition.** Given an  $m \times n$  matrix  $\mathbf{A}$ , its transpose  $\mathbf{A}^T$  is the  $n \times m$  matrix such that

$$(\mathbf{A}^T)_{ij} = \mathbf{A}_{ji}.$$

**Example.** Consider

$$\mathbf{M} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6\pi \end{bmatrix}.$$

**Theorem 3.** Let  $\mathbf{A}$  and  $\mathbf{B}$  be matrices whose sizes are appropriate for the following sums and products. Then

1.  $(\mathbf{A}^T)^T = \mathbf{A}$
2.  $(\mathbf{A} + \mathbf{B})^T = \mathbf{A}^T + \mathbf{B}^T$
3.  $(r\mathbf{A})^T = r\mathbf{A}^T$
4.  $(\mathbf{AB})^T = \mathbf{B}^T\mathbf{A}^T$

Matrix inverses

Invertible square matrices are more well behaved than arbitrary square matrices.

**Definition.** Let  $\mathbf{A}$  be a square matrix such that there exists a square matrix  $\mathbf{B}$  such that either:

1.  $\mathbf{AB} = \mathbf{I}$  or
2.  $\mathbf{BA} = \mathbf{I}$ .

Then we say that  $\mathbf{A}$  is *invertible* and that  $\mathbf{B}$  is the *inverse* of  $\mathbf{A}$ .

**Examples.** Consider

$$\mathbf{A}_1 = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \quad \text{and} \quad \mathbf{A}_2 = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}.$$