

## A few other bits of matrix notation and facts

$\mathbf{AB} \neq \mathbf{BA}$ . Other exceptions are associated with zero matrices. A **zero matrix** is one whose elements are *all* zero, such as

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

We ordinarily denote a zero matrix (whatever its size) by  $\mathbf{0}$ . It should be clear that for any matrix  $\mathbf{A}$ ,

$$\mathbf{0} + \mathbf{A} = \mathbf{A} = \mathbf{A} + \mathbf{0}, \quad \mathbf{A}\mathbf{0} = \mathbf{0}, \quad \text{and} \quad \mathbf{0}\mathbf{A} = \mathbf{0},$$

where in each case  $\mathbf{0}$  is a zero matrix of appropriate size. Thus zero matrices appear to play a role in the arithmetic of matrices similar to the role of the real number 0 in ordinary arithmetic.

Recall that an *identity matrix* is a square matrix  $\mathbf{I}$  that has ones on its principal diagonal and zeros elsewhere. Identity matrices play a role in matrix arithmetic which is strongly analogous to that of the real number 1, for which  $a \cdot 1 = 1 \cdot a = a$  for all values of the real number  $a$ . For instance, you can check that

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}.$$

Similarly, if

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \quad \text{and} \quad \mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix},$$

then  $\mathbf{AI} = \mathbf{IA} = \mathbf{A}$ . For instance, the element in the second row and third column of  $\mathbf{AI}$  is

$$(a_{21})(0) + (a_{22})(0) + (a_{23})(1) = a_{23}.$$

Recall that the  $n \times n$  **identity matrix** is the diagonal matrix

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{bmatrix} \quad (1)$$

having ones on its main diagonal and zeros elsewhere. It is not difficult to deduce directly from the definition of the matrix product that  $\mathbf{I}$  acts like an identity for matrix multiplication:

$$\mathbf{AI} = \mathbf{A} \quad \text{and} \quad \mathbf{IB} = \mathbf{B} \quad (2)$$

if the sizes of  $\mathbf{A}$  and  $\mathbf{B}$  are such that the products  $\mathbf{AI}$  and  $\mathbf{IB}$  are defined. It is, nevertheless, instructive to derive the identities in (2) formally from the two basic facts about matrix multiplication that we state below. First, recall that the notation

$$\mathbf{A} = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \cdots \quad \mathbf{a}_n] \quad (3)$$

expresses the  $m \times n$  matrix  $\mathbf{A}$  in terms of its column vectors  $\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3, \dots, \mathbf{a}_n$ .

**Fact 1**  $\mathbf{Ax}$  in terms of columns of  $\mathbf{A}$

If  $\mathbf{A} = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n]$  and  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  is an  $n$ -vector, then

$$\mathbf{Ax} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n. \quad (4)$$

The reason is that when each row vector of  $\mathbf{A}$  is multiplied by the column vector  $\mathbf{x}$ , its  $j$ th element is multiplied by  $x_j$ .

**Fact 2**  $\mathbf{AB}$  in terms of columns of  $\mathbf{B}$

If  $\mathbf{A}$  is an  $m \times n$  matrix and  $\mathbf{B} = [\mathbf{b}_1 \quad \mathbf{b}_2 \quad \cdots \quad \mathbf{b}_p]$  is an  $n \times p$  matrix, then

$$\mathbf{AB} = [\mathbf{Ab}_1 \quad \mathbf{Ab}_2 \quad \cdots \quad \mathbf{Ab}_p]. \quad (5)$$

That is, *the  $j$ th column of  $\mathbf{AB}$  is the product of  $\mathbf{A}$  and the  $j$ th column of  $\mathbf{B}$* . The reason is that the elements of the  $j$ th column of  $\mathbf{AB}$  are obtained by multiplying the individual rows of  $\mathbf{A}$  by the  $j$ th column of  $\mathbf{B}$ .

**Example 1** The third column of the product  $\mathbf{AB}$  of the matrices

$$\mathbf{A} = \begin{bmatrix} 2 & -1 & 0 \\ 4 & 0 & 3 \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} 3 & 7 & 5 & -4 \\ -2 & 6 & 3 & 6 \\ 5 & 1 & -2 & -1 \end{bmatrix}$$

is

$$\mathbf{Ab}_3 = \begin{bmatrix} 2 & -1 & 0 \\ 4 & 0 & 3 \end{bmatrix} \begin{bmatrix} 5 \\ 3 \\ -2 \end{bmatrix} = \begin{bmatrix} 7 \\ 14 \end{bmatrix}. \quad \blacksquare$$

To prove that  $\mathbf{AI} = \mathbf{A}$ , note first that

$$\mathbf{I} = [\mathbf{e}_1 \quad \mathbf{e}_2 \quad \cdots \quad \mathbf{e}_n], \quad (6)$$

where the  $j$ th column vector of  $\mathbf{I}$  is the  $j$ th **basic unit vector**

$$\mathbf{e}_j = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \leftarrow j \text{th entry.} \quad (7)$$

If  $\mathbf{A} = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n]$ , then Fact 1 yields

$$\mathbf{Ae}_j = 0 \cdot \mathbf{a}_1 + \cdots + 1 \cdot \mathbf{a}_j + \cdots + 0 \cdot \mathbf{a}_n = \mathbf{a}_j. \quad (8)$$

Hence Fact 2 gives

$$\begin{aligned} \mathbf{AI} &= \mathbf{A} [\mathbf{e}_1 \quad \mathbf{e}_2 \quad \cdots \quad \mathbf{e}_n] \\ &= [\mathbf{Ae}_1 \quad \mathbf{Ae}_2 \quad \cdots \quad \mathbf{Ae}_n] = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n]; \end{aligned}$$

that is,  $\mathbf{AI} = \mathbf{A}$ . The proof that  $\mathbf{IB} = \mathbf{B}$  is similar. (See Problems 41 and 42.)