These first two pages are reminders of parts of section 3.5 that I had previously indicated you all ought to read. They are relevant for the stuff below on matrix equations.

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

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

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

$$AI = A$$
 and $IB = B$ (2)

if the sizes of **A** and **B** are such that the products **AI** and **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

$$\mathbf{A} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 & \cdots & \mathbf{a}_n \end{bmatrix} \tag{3}$$

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

Fact 1 Ax in terms of columns of A

If
$$\mathbf{A} = [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n]$$
 and $\mathbf{x} = (x_1, x_2, \dots, x_n)$ is an *n*-vector, then
$$\mathbf{A}\mathbf{x} = x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_n \mathbf{a}_n. \tag{4}$$

The reason is that when each row vector of **A** is multiplied by the column vector **x**, its *j* th element is multiplied by x_j .

Fact 2 AB in terms of columns of B

If **A** is an $m \times n$ matrix and **B** = $\begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \cdots & \mathbf{b}_p \end{bmatrix}$ is an $n \times p$ matrix, then

$$\mathbf{AB} = \begin{bmatrix} \mathbf{Ab}_1 & \mathbf{Ab}_2 & \cdots & \mathbf{Ab}_p \end{bmatrix}. \tag{5}$$

That is, the jth column of AB is the product of A and the jth column of B. The reason is that the elements of the jth column of AB are obtained by multiplying the individual rows of A by the jth column of B.

The third column of the product 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

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

To prove that AI = A, note first that

$$\mathbf{I} = \begin{bmatrix} \mathbf{e}_1 & \mathbf{e}_2 & \cdots & \mathbf{e}_n \end{bmatrix}, \tag{6}$$

where the jth column vector of **I** is the jth **basic unit vector**

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

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

$$\mathbf{A}\mathbf{e}_{j} = 0 \cdot \mathbf{a}_{1} + \dots + 1 \cdot \mathbf{a}_{j} + \dots + 0 \cdot \mathbf{a}_{n} = \mathbf{a}_{j}. \tag{8}$$

Hence Fact 2 gives

$$\mathbf{AI} = \mathbf{A} \begin{bmatrix} \mathbf{e}_1 & \mathbf{e}_2 & \cdots & \mathbf{e}_n \end{bmatrix}$$
$$= \begin{bmatrix} \mathbf{Ae}_1 & \mathbf{Ae}_2 & \cdots & \mathbf{Ae}_n \end{bmatrix} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \cdots & \mathbf{a}_n \end{bmatrix};$$

that is, AI = A. The proof that IB = B is similar. (See Problems 41 and 42.)

Matrix Equations

In certain applications, one needs to solve a system $A\mathbf{x} = \mathbf{b}$ of n equations in n unknowns several times in succession—with the same $n \times n$ coefficient matrix \mathbf{A} each time, but with different constant vectors $\mathbf{b}_1, \mathbf{b}_2, \ldots, \mathbf{b}_k$ on the right. Thus we want to find solution vectors $\mathbf{x}_1, \mathbf{x}_2, \ldots, \mathbf{x}_k$ such that

$$\mathbf{A}\mathbf{x}_1 = \mathbf{b}_1, \quad \mathbf{A}\mathbf{x}_2 = \mathbf{b}_2, \quad \dots, \quad \mathbf{A}\mathbf{x}_k = \mathbf{b}_k. \tag{17}$$

By Fact 2 at the beginning of this section,

$$[\mathbf{A}\mathbf{x}_1 \quad \mathbf{A}\mathbf{x}_2 \quad \cdots \quad \mathbf{A}\mathbf{x}_k] = \mathbf{A}[\mathbf{x}_1 \quad \mathbf{x}_2 \quad \cdots \quad \mathbf{x}_k].$$

So the k equations in (17) are equivalent to the single matrix equation

$$\mathbf{AX} = \mathbf{B},\tag{18}$$

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where

$$\mathbf{X} = [\mathbf{x}_1 \ \mathbf{x}_2 \ \cdots \ \mathbf{x}_k]$$
 and $\mathbf{B} = [\mathbf{b}_1 \ \mathbf{b}_2 \ \cdots \ \mathbf{b}_k]$.

If **A** is invertible and we know A^{-1} , we can find the $n \times k$ matrix of "unknowns" by multiplying each term in Equation (18) on the left by A^{-1} :

$$\mathbf{X} = \mathbf{A}^{-1}\mathbf{B}.\tag{19}$$

Note that this equation is a generalization of Eq. (13) in Theorem 4. If k = 1, it usually is simplest to solve the system by Gaussian elimination, but when several different solutions are sought, it may be simpler to find A^{-1} first and then to apply (19).

Continue to next page.

Find a 3×4 matrix **X** such that

$$\begin{bmatrix} 4 & 3 & 2 \\ 5 & 6 & 3 \\ 3 & 5 & 2 \end{bmatrix} \mathbf{X} = \begin{bmatrix} 3 & -1 & 2 & 6 \\ 7 & 4 & 1 & 5 \\ 5 & 2 & 4 & 1 \end{bmatrix}.$$

The coefficient matrix is the matrix **A** whose inverse we found in Example 7, so Eq. (19) yields

$$\mathbf{X} = \mathbf{A}^{-1}\mathbf{B} = \begin{bmatrix} 3 & -4 & 3 \\ 1 & -2 & 2 \\ -7 & 11 & -9 \end{bmatrix} \begin{bmatrix} 3 & -1 & 2 & 6 \\ 7 & 4 & 1 & 5 \\ 5 & 2 & 4 & 1 \end{bmatrix},$$

and hence

$$\mathbf{X} = \begin{bmatrix} -4 & -13 & 14 & 1 \\ -1 & -5 & 8 & -2 \\ 11 & 33 & -39 & 4 \end{bmatrix}.$$

By looking at the third columns of B and X, for instance, we see that the solution of

$$4x_1 + 3x_2 + 2x_3 = 2$$

$$5x_1 + 6x_2 + 3x_3 = 1$$

$$3x_1 + 5x_2 + 2x_3 = 4$$

is
$$x_1 = 14$$
, $x_2 = 8$, $x_3 = -39$.