

# APPENDIX A

## Partitioned Matrices and Kronecker Products

### Partitioned Matrices

It is sometimes useful to partition the elements of a matrix into  $MN$  submatrices as

$$\mathbf{A} = \begin{bmatrix} \mathbf{A}_{11} & \cdots & \mathbf{A}_{1N} \\ \vdots & & \vdots \\ \mathbf{A}_{M1} & \cdots & \mathbf{A}_{MN} \end{bmatrix}.$$

This is a **partitioned matrix**. The subscript of the submatrices are defined in the same fashion as those for the elements of a matrix. For example, we might write

$$\mathbf{A} = \begin{bmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{bmatrix} = \begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{bmatrix},$$

with

$$\mathbf{A}_{11} = \begin{bmatrix} 1 & 4 \\ 2 & 5 \end{bmatrix}, \quad \mathbf{A}_{12} = \begin{bmatrix} 7 \\ 8 \end{bmatrix}, \quad \mathbf{A}_{21} = [3 \ 6], \quad \mathbf{A}_{22} = 9.$$

A common special case is where  $M = N$  and the off-diagonal blocks ( $\mathbf{A}_{mn}$  for  $m \neq n$ ) are all zero matrices:

$$\mathbf{A} = \begin{bmatrix} \mathbf{A}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM} \end{bmatrix}.$$

This is a **block diagonal matrix**.

### Addition and Multiplication of Partitioned Matrices

Matrix addition and multiplication extend to partitioned matrices. Therefore,

$$\mathbf{A} + \mathbf{B} = \begin{bmatrix} \mathbf{A}_{11} + \mathbf{B}_{11} & \cdots & \mathbf{A}_{1N} + \mathbf{B}_{1N} \\ \vdots & & \vdots \\ \mathbf{A}_{M1} + \mathbf{B}_{M1} & \cdots & \mathbf{A}_{MN} + \mathbf{B}_{MN} \end{bmatrix}, \quad (\text{A.1})$$

$$\begin{aligned} \mathbf{AB} &= \begin{bmatrix} \mathbf{A}_{11} & \cdots & \mathbf{A}_{1N} \\ \vdots & & \vdots \\ \mathbf{A}_{M1} & \cdots & \mathbf{A}_{MN} \end{bmatrix} \begin{bmatrix} \mathbf{B}_{11} & \cdots & \mathbf{B}_{1L} \\ \vdots & & \vdots \\ \mathbf{B}_{N1} & \cdots & \mathbf{B}_{NL} \end{bmatrix} \\ &= \begin{bmatrix} \sum_{n=1}^N \mathbf{A}_{1n} \mathbf{B}_{n1} & \cdots & \sum_{n=1}^N \mathbf{A}_{1n} \mathbf{B}_{nL} \\ \vdots & & \vdots \\ \sum_{n=1}^N \mathbf{A}_{Mn} \mathbf{B}_{n1} & \cdots & \sum_{n=1}^N \mathbf{A}_{Mn} \mathbf{B}_{nL} \end{bmatrix}. \end{aligned} \quad (\text{A.2})$$

In all these expressions, the matrices must be conformable for the operations involved. With respect to addition, the dimension of  $\mathbf{A}_{mn}$  and  $\mathbf{B}_{mn}$  must be the same for all  $m (= 1, 2, \dots, M)$  and  $n (= 1, 2, \dots, N)$ . For multiplication, the number of columns in  $\mathbf{A}_{mn}$  must equal the number of rows in  $\mathbf{B}_{n\ell}$  for all  $m (= 1, 2, \dots, M)$ ,  $n (= 1, 2, \dots, N)$ , and  $\ell (= 1, 2, \dots, L)$ . A special case of multiplication is when  $\mathbf{B}$  is a **stacked vector c**:

$$\mathbf{Ac} = \begin{bmatrix} \mathbf{A}_{11} & \cdots & \mathbf{A}_{1N} \\ \vdots & & \vdots \\ \mathbf{A}_{M1} & \cdots & \mathbf{A}_{MN} \end{bmatrix} \begin{bmatrix} \mathbf{c}_1 \\ \vdots \\ \mathbf{c}_N \end{bmatrix} = \begin{bmatrix} \sum_{n=1}^N \mathbf{A}_{1n} \mathbf{c}_n \\ \vdots \\ \sum_{n=1}^N \mathbf{A}_{Mn} \mathbf{c}_n \end{bmatrix}. \quad (\text{A.3})$$

Several cases frequently encountered are of the form

$$\begin{bmatrix} \mathbf{A}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{c}_1 \\ \vdots \\ \mathbf{c}_M \end{bmatrix} = \begin{bmatrix} \mathbf{A}_{11} \mathbf{c}_1 \\ \vdots \\ \mathbf{A}_{MM} \mathbf{c}_M \end{bmatrix}, \quad (\text{A.4})$$



$$\begin{bmatrix} \mathbf{A}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{B}_{11} & & \\ & \ddots & \\ & & \mathbf{B}_{MM} \end{bmatrix} = \begin{bmatrix} \mathbf{A}_{11}\mathbf{B}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM}\mathbf{B}_{MM} \end{bmatrix}, \quad (\text{A.5})$$

$$\begin{bmatrix} \mathbf{A}'_{11} & & \\ & \ddots & \\ & & \mathbf{A}'_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{B}_{11} & \cdots & \mathbf{B}_{1M} \\ \vdots & & \vdots \\ \mathbf{B}_{M1} & \cdots & \mathbf{B}_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{A}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM} \end{bmatrix} \\ = \begin{bmatrix} \mathbf{A}'_{11}\mathbf{B}_{11}\mathbf{A}_{11} & \cdots & \mathbf{A}'_{11}\mathbf{B}_{1M}\mathbf{A}_{MM} \\ \vdots & & \vdots \\ \mathbf{A}'_{MM}\mathbf{B}_{M1}\mathbf{A}_{11} & \cdots & \mathbf{A}'_{MM}\mathbf{B}_{MM}\mathbf{A}_{MM} \end{bmatrix}, \quad (\text{A.6})$$

$$\begin{bmatrix} \mathbf{A}'_{11} & & \\ & \ddots & \\ & & \mathbf{A}'_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{B}_{11} & \cdots & \mathbf{B}_{1M} \\ \vdots & & \vdots \\ \mathbf{B}_{M1} & \cdots & \mathbf{B}_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{c}_1 \\ \vdots \\ \mathbf{c}_M \end{bmatrix} \\ = \begin{bmatrix} \mathbf{A}'_{11}\mathbf{B}_{11}\mathbf{c}_1 + \cdots + \mathbf{A}'_{11}\mathbf{B}_{1M}\mathbf{c}_M \\ \vdots \\ \mathbf{A}'_{MM}\mathbf{B}_{M1}\mathbf{c}_1 + \cdots + \mathbf{A}'_{MM}\mathbf{B}_{MM}\mathbf{c}_M \end{bmatrix}, \quad (\text{A.7})$$

$$\begin{bmatrix} \mathbf{A}'_{11} & \cdots & \mathbf{A}'_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{B}_{11} & \cdots & \mathbf{B}_{1M} \\ \vdots & & \vdots \\ \mathbf{B}_{M1} & \cdots & \mathbf{B}_{MM} \end{bmatrix} \begin{bmatrix} \mathbf{A}_{11} \\ \vdots \\ \mathbf{A}_{MM} \end{bmatrix} = \sum_{m=1}^M \sum_{h=1}^M \mathbf{A}'_{mm} \mathbf{B}_{mh} \mathbf{A}_{hh}, \quad (\text{A.8})$$

where  $\mathbf{c}_m$  ( $m = 1, 2, \dots, M$ ) are column vectors.

### Inverting Partitioned Matrices

The inverse of a block diagonal matrix is

$$\begin{bmatrix} \mathbf{A}_{11} & & \\ & \ddots & \\ & & \mathbf{A}_{MM} \end{bmatrix}^{-1} = \begin{bmatrix} \mathbf{A}_{11}^{-1} & & \\ & \ddots & \\ & & \mathbf{A}_{MM}^{-1} \end{bmatrix}, \quad (\text{A.9})$$

provided  $\mathbf{A}_{mm}$  ( $m = 1, 2, \dots, M$ ) are invertible. This can be verified by direct multiplication.

For the general  $2 \times 2$  partitioned matrix, one form of the **partitioned inverse** is

$$\begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{bmatrix}^{-1} = \begin{bmatrix} \mathbf{A}_{11}^{-1} + \mathbf{A}_{11}^{-1}\mathbf{A}_{12}\mathbf{F}_2\mathbf{A}_{21}\mathbf{A}_{11}^{-1} & -\mathbf{A}_{11}^{-1}\mathbf{A}_{12}\mathbf{F}_2 \\ -\mathbf{F}_2\mathbf{A}_{21}\mathbf{A}_{11}^{-1} & \mathbf{F}_2 \end{bmatrix}, \quad (\text{A.10})$$

where  $\mathbf{F}_2 = (\mathbf{A}_{22} - \mathbf{A}_{21}\mathbf{A}_{11}^{-1}\mathbf{A}_{12})^{-1}$ . This can be verified easily by multiplying  $\mathbf{A}$  by the inverse. In view of the symmetry of the calculation, the upper left block can also be written as

$$\mathbf{F}_1 = (\mathbf{A}_{11} - \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{A}_{21})^{-1}.$$

### Kronecker Products

For general matrices  $\mathbf{A}$  ( $M \times N$ ) and  $\mathbf{B}$  ( $K \times L$ ), the **Kronecker product** is defined as

$$\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} a_{11}\mathbf{B} & \cdots & a_{1N}\mathbf{B} \\ \vdots & & \vdots \\ a_{M1}\mathbf{B} & \cdots & a_{MN}\mathbf{B} \end{bmatrix}. \quad (\text{A.11})$$

This is an  $MK \times NL$  matrix. A special case is when both  $\mathbf{A}$  and  $\mathbf{B}$  are vectors:

$$\mathbf{a} \otimes \mathbf{b} = \begin{bmatrix} a_1 \\ \vdots \\ a_M \end{bmatrix} \otimes \begin{bmatrix} b_1 \\ \vdots \\ b_N \end{bmatrix} = \begin{bmatrix} a_1\mathbf{b} \\ \vdots \\ a_M\mathbf{b} \end{bmatrix}. \quad (\text{A.12})$$

It is straightforward but cumbersome to verify the following:

$$(\mathbf{A} \otimes \mathbf{B})(\mathbf{C} \otimes \mathbf{D}) = \mathbf{AC} \otimes \mathbf{BD}, \quad (\text{A.13})$$

provided  $\mathbf{A}$  and  $\mathbf{C}$  are conformable and  $\mathbf{B}$  and  $\mathbf{D}$  are conformable, and

$$(\mathbf{A} \otimes \mathbf{B})' = \mathbf{A}' \otimes \mathbf{B}', \quad (\text{A.14})$$

$$(\mathbf{A} \otimes \mathbf{B})^{-1} = \mathbf{A}^{-1} \otimes \mathbf{B}^{-1}, \quad (\text{A.15})$$

provided  $\mathbf{A}$  and  $\mathbf{B}$  are invertible.