Matrix Algebra

Matrices

A matrix is a rectangular array of elements arranged in rows and columns, e.g.:

$$\mathbf{A} = \begin{bmatrix} 6 & 13 \\ 9 & 21 \\ 12 & 5 \end{bmatrix}, \text{ where } \mathbf{A} \text{ here is a 3 row by 2 column matrix.}$$

More generally, a matrix of this size or dimension (3 by 2) can be written as

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}$$
 where the a_{ij} 's are the elements of the matrix.

Even more generally, the $(r \times c)$ matrix **A** can be written as:

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1j} & \cdots & a_{1c} \\ a_{21} & a_{22} & \cdots & a_{2j} & \cdots & a_{2c} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{i1} & a_{i2} & \cdots & a_{ij} & \cdots & a_{ic} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{r1} & a_{r2} & \cdots & a_{rj} & \cdots & a_{rc} \end{bmatrix}$$

where $\mathbf{A} = [a_{ij}]$; i = 1, ..., r; j = 1, ..., c; and *r* is the number of rows and *c* the number of columns in the matrix.

Special Matrices

1) the square matrix (r = c), e.g.

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix};$$

2) the 1-d row vector (an $r \times 1$ matrix), e.g.

$$\mathbf{a} = \begin{bmatrix} a_{11} \\ a_{21} \\ \cdots \\ a_{r1} \end{bmatrix};$$

3) the 1-d *column vector* (a $1 \times c$ matrix), e.g.

$$\mathbf{b} = \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1c} \end{bmatrix};$$

4) a *scalar*, or a 1×1 matrix: $\mathbf{a} = [a_{11}]$

5) the *zero matrix*:

$$\mathbf{0} = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & 0 \end{bmatrix};$$

6) an $r \times r$ diagonal matrix:

$$\mathbf{T} = \begin{bmatrix} t_1 & 0 & \cdots & 0 & 0 \\ 0 & t_2 & \cdots & 0 & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & t_{r-1} & 0 \\ 0 & 0 & \cdots & 0 & t_r \end{bmatrix};$$

7) the *identity matrix*:

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \cdots & \cdots & \cdots & 0 \\ 0 & 0 & \cdots & 1 \end{bmatrix}; \text{ (a square matrix of zeros, with ones along the diagonal), and}$$

8) a vector of ones: $\mathbf{1} = \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}$.

Matrix operations and further definitions:

1) transposition, e.g., if

$$\mathbf{A} = \begin{bmatrix} 1 & 6 \\ 5 & 3 \\ 0 & 2 \end{bmatrix}, \text{ then } \mathbf{A}^T \text{ or } \mathbf{A}' = \begin{bmatrix} 1 & 5 & 0 \\ 6 & 3 & 2 \end{bmatrix}, \text{ or if}$$
$$\mathbf{C} = \begin{bmatrix} 7 \\ 9 \\ 5 \end{bmatrix}, \text{ then } \mathbf{C}' = \begin{bmatrix} 7 & 9 & 5 \end{bmatrix}, \text{ or more generally, if } \mathbf{A} \text{ (an } r \times c \text{ matrix) is}$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & \cdots & a_{1c} \\ \cdots & \cdots & \cdots \\ a_{r1} & \cdots & a_{rc} \end{bmatrix}, \text{ then } \mathbf{A}' = \begin{bmatrix} a_{11} & \cdots & a_{r1} \\ \cdots & \cdots & \cdots \\ a_{1c} & \cdots & a_{rc} \end{bmatrix}.$$

2) a matrix can be said to be *symmetric* if A = A';

3) equality: $\mathbf{A} = \mathbf{B}$, if $a_{ij} = b_{ij}$, all *i* and *j*.

4) addition and subtraction (of corresponding elements): If

$$\mathbf{A} = \begin{bmatrix} 1 & 4 \\ 2 & 3 \\ 3 & 6 \end{bmatrix} \text{ and } \mathbf{B} = \begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \end{bmatrix}, \text{ then}$$
$$\mathbf{A} + \mathbf{B} = \begin{bmatrix} 1+1 & 4+2 \\ 2+2 & 5+3 \\ 3+3 & 6+4 \end{bmatrix} = \begin{bmatrix} 2 & 6 \\ 4 & 8 \\ 6 & 10 \end{bmatrix}, \text{ and } \mathbf{A} - \mathbf{B} = \begin{bmatrix} 1-1 & 4-2 \\ 2-2 & 5-3 \\ 3-3 & 6-4 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 0 & 2 \\ 0 & 2 \end{bmatrix}.$$

More generally, $\mathbf{C} = \mathbf{A} + \mathbf{B}$, where $c_{ij} = a_{ij} + b_{ij}$.

In order to be added or subtracted, the two matrices must be *conformable*. In the case of addition or subtraction, they must be of the same shape. If **A** is a *p* row by *q* column matrix, which can be written as ${}_{p}\mathbf{A}_{q}$, then **B** must also be a *p* row by *q* column matrix, and the resultant matrix **C** will also be a *p* row by *q* column matrix, or ${}_{p}\mathbf{C}_{q} = {}_{p}\mathbf{A}_{q} + {}_{p}\mathbf{B}_{q}$.

5) scalar multiplication (multiplication of each element by the same scalar value): if

$$\mathbf{A} = \begin{bmatrix} 2 & 7 \\ 9 & 3 \end{bmatrix}, \text{ then } 4 \times \mathbf{A} = 4 \cdot \begin{bmatrix} 2 & 7 \\ 9 & 3 \end{bmatrix} = \begin{bmatrix} 8 & 28 \\ 36 & 12 \end{bmatrix}.$$

6) matrix multiplication: if

$$\mathbf{A} = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}, \text{ and } \mathbf{B} = \begin{bmatrix} 2 & 3 \\ 0 & 4 \end{bmatrix}, \text{ then}$$
$$\mathbf{C} = \mathbf{A}\mathbf{B} = \begin{bmatrix} (1 \cdot 2 + 3 \cdot 0) & (1 \cdot 3 + 3 \cdot 4) \\ (2 \cdot 2 + 4 \cdot 0) & (3 \cdot 3 + 4 \cdot 4) \end{bmatrix}.$$

Note that matrix multiplication requires that the matrices be *conformable* for multiplication, e.g. if **A** is a *p* row by *q* column matrix, and **B** is a *q* row by *r* column matrix then ${}_{p}\mathbf{A}_{q}$ and ${}_{q}\mathbf{B}_{r}$ can be said to be conformable (because **A** has *q* columns, and **B** has *q* rows), and their product **C** can be obtained with dimensions of *p* rows by *r* columns, or ${}_{p}\mathbf{C}_{r} = {}_{p}\mathbf{A}_{q}\mathbf{B}_{r}$.

Note that **A** postmultiplied by **B** is not conformal, except if p = r.

Matrix multiplication can also be considered as a set of sums of cross-products, one for each element of the product matrix, or

$$_{p}\mathbf{C}_{r} = {}_{p}\mathbf{A}_{q}\mathbf{B}_{r}, \text{ where } c_{ij} = \sum_{k=1}^{q} a_{ik}b_{jk}.$$

7) matrix *inversion*: For a square matrix **A**, the inverse of **A** is the matrix that when premultiplied or postmultiplied by **A** yields the identity matix **I**, or

$$\mathbf{A}^{-1}\mathbf{A} = \mathbf{A}\mathbf{A}^{-1} = \mathbf{I}$$

Matrix inversion is analogous in some ways to scalar division:

If
$$(1/x)$$
 is the inverse of x, then $x \cdot \frac{1}{x} = xx^{-1} = x^{-1}x = 1$.

8) Linear combinations can be compactly written in matix terms. If

$$z_{1} = a_{1}X_{11} + a_{2}X_{12} + \dots + a_{p}X_{1p}$$

$$z_{2} = a_{1}X_{21} + a_{2}X_{22} + \dots + a_{p}X_{2p}$$

$$\dots$$

$$z_{i} = a_{1}X_{i1} + a_{2}X_{i2} + \dots + a_{p}X_{ip}$$

$$\dots$$

$$z_{n} = a_{1}X_{n1} + a_{2}X_{n2} + \dots + a_{p}X_{np}$$

then this system of equations can be written in more compact form as

$$\mathbf{z}_{(N\times1)} = \mathbf{X}_{(N\timesp)(p\times1)} \mathbf{b}, \text{ where}$$
$$\mathbf{z}_{(N\times1)} = \begin{bmatrix} z_1 \\ z_2 \\ \cdots \\ z_N \end{bmatrix}, \mathbf{a}_{(p\times1)} = \begin{bmatrix} a_0 \\ a_1 \\ \cdots \\ a_p \end{bmatrix}, \text{ and } \mathbf{X}_{(N\timesp)} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1p} \\ x_{21} & x_{22} & \cdots & x_{2p} \\ \cdots & \cdots & \cdots & \cdots \\ x_{N1} & x_{N2} & \cdots & x_{Np} \end{bmatrix}.$$

9) *Eigenvalues* and *eigenvectors*. If **R** is a $(p \times p)$ square matrix, then there is another $(p \times p)$ square matrix **E**, such that

RE = **EV**, where **V** is a $(p \times p)$ diagonal matrix.

The first column of these matrices can be written as $\mathbf{R}\mathbf{e}_1 = \lambda_1 \mathbf{e}_1$, or $(\mathbf{R} - \lambda_1 \mathbf{I})\mathbf{e}_1 = \mathbf{0}$.

10) Matrix quadratic forms: If A is an $(n \times n)$ and x is an $(n \times 1)$ column vector, then

$$Q = \mathbf{x}' \mathbf{A} \mathbf{x} = \mathbf{x} \mathbf{A} \mathbf{x}' = \sum_{i=1}^{n} \sum_{j=1}^{n} x_i a_{ij} x_j.$$