# Matrix algebra

and

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#### Section Overview

# • Key Points: Matrix algebra

- Definition of a matrix
- Matrix addition and scalar multiplication
- Matrix multiplication
- Matrix powers
- Transpose of a matrix
- Inverse of a matrix
- Determinant of a matrix

#### • Associated sections of the book:

- Poole Section 3.1 (p138 152) (Omit partitioned matrices p145-149)
- Poole Section 3.2 (P160) (Omit column operations)
- Poole Section 3.3 (P169)

#### 3.1: Matrix Operations

**Definition** A matrix is a rectangular array of numbers called the entries or elements of the matrix.

For a matrix 
$$A$$
 we write the  $i$ ,  $j^{th}$  entry as  $a_{ij}$ .  
For the matrix  $A = \begin{bmatrix} 2 & 0 & 5 \\ 1 & 4 & -1 \end{bmatrix}$  then  $a_{11} = 2$ ,  $a_{23} = -1$ ,  $a_{21} = 1$  and  $a_{12} = 0$ .

Symbolically this is expressed by  $A = (a_{ij})$ . So if A is an  $m \times n$ matrix then

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}.$$

If the columns of *A* are the column vectors  $\mathbf{a}_1$ ,  $\mathbf{a}_2$ ,  $\cdots$ ,  $\mathbf{a}_n$  in  $\mathbb{R}^m$  we write

$$A = [\mathbf{a}_1 \cdots \mathbf{a}_n],$$

and if the rows are row vectors  $\mathbf{b}_1$ ,  $\mathbf{b}_2$ ,  $\cdots$ ,  $\mathbf{b}_m$  in  $\mathbb{R}^n$  we write

$$A = \left[ egin{array}{c} \mathbf{b}_1 \ dots \ \mathbf{b}_m \end{array} 
ight].$$

A square matrix (m = n) is diagonal iff its off-diagonal elements are zero. The  $n \times n$  identity matrix  $I_n$  is the diagonal matrix with all diagonal entries 1.

Example Consider the matrices 
$$A = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 4 & 4 \\ 0 & 0 & 8 \end{bmatrix}$$
,  $B = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$  and  $C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

Two matrices are equal if and only if they have the same size and their corresponding entries are equal.

#### Matrix addition and scalar multiplication

If  $A = (a_{ij})$  and  $B = (b_{ij})$  are  $m \times n$  matrices and c is a scalar <sup>2</sup> and *c* is a scalar, then we can define new  $m \times n$  matrices A + B and cA componentwise: <sup>3</sup>

$$A + B = (a_{ij} + b_{ij}),$$
  

$$cA = c(a_{ij}) = (ca_{ij}).$$

#### Matrix multiplication

If *A* is an  $m \times n$  matrix and *B* is an  $n \times r$  matrix then C = AB is the  $m \times r$  matrix with  $(i, j)^{th}$  entry given by<sup>4</sup>

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj} = \sum_{k=1}^{n} a_{ik}b_{kj}.$$

Theorem **3.1** Let A be an  $m \times n$  matrix, and

$$\mathbf{e}_i = [0, \cdots, 0, 1, 0, \cdots, 0],$$

- <sup>2</sup> Note that matrices must be the same size for addition of two matrices to be defined.
- <sup>3</sup> Note that we have the same two operations: addition and scalar multiplication defined for  $m \times n$  matrices as we had in  $\mathbb{R}^n$ . These operations satisfy the same rules as our vectors did in Theorem 1.1 (see Theorem 3.2 below). So, the collection of  $m \times n$  matrices is a vector space in the sense that we will define in chapter 6. We're starting to see the potential value in generalisation: the additive structure of matrices and of vectors is the same: we can (and will) prove general theorems which will simultaneously cover both situations.
- <sup>4</sup> Again note that matrix multiplication is not defined for every pair (A, B) of matrices. It is only defined when the number of rows of A is equal to the number of columns of B. Mathematics is heavily typed: operations can only be performed when objects have the right type (which of course depends on the operation).

with the 1 in the  $i^{th}$  position, and

$$\mathbf{e}_{j} = \left[ egin{array}{c} 0 \\ dots \\ 0 \\ 1 \\ 0 \\ dots \\ 0 \end{array} 
ight],$$

with the 1 in the  $j^{th}$  position. Then

- (a)  $\mathbf{e}_i A$  is the  $i^{\text{th}}$  row of A,
- (b)  $A\mathbf{e}_j$  is the  $j^{\text{th}}$  column of A.

Proof: Omitted.

#### **Matrix Powers**

If *A* is an  $n \times n$  matrix and *k* is a positive integer then

$$A^0 = \mathbb{I}_n,$$
  $A^2 = AA,$   $A^k = \underbrace{AA \cdots A}_{k \text{ times}}.$ 

Example If 
$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$
 then compute  $A^2$  and  $A^3$ .

### The transpose of a matrix

**Definition** The *transpose* of an  $m \times n$  matrix A is the  $n \times m$  matrix  $A^{T}$  obtained by interchanging the rows and columns of A.

**Definition** A square matrix is *symmetric* if  $A^{T} = A$ . That is, A is equal to its own transpose.

Example Are the matrices  $A = \begin{bmatrix} 2 & 0 \\ 1 & 3 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 3 & 7 \\ 1 & 7 & 5 \end{bmatrix}$  symmetric?

# 3.2: Matrix Algebra

The addition and scalar multiplication rules for  $m \times n$  matrices obey the following algebraic properties.<sup>5</sup>

**Theorem** 3.2 Let A, B and C be  $m \times n$  matricies and c and d be scalars. Then

(a) 
$$A + B = B + A$$
,

(b) 
$$(A+B)+C=A+(B+C)$$
,

(c) 
$$A + 0 = A$$
,

<sup>5</sup> Note the similarity between this and Theorem 1.1.

- (d) A + (-A) = 0,
- (e) c(A + B) = cA + cB,
- (f) (c+d)A = cA + dA,
- (g) c(dA) = (cd)A,
- (h) 1A = A.

Proof:

# Properties of matrix multiplication

Matrix multiplication behaves differently from multiplication of numbers. In general multiplication is not commutative. 6 Also, we could have  $A^2 = 0$  even if  $A \neq 0.7$ 

**Theorem** 3.3 Let A, B and C be matrices and k be a scalar. The following identities hold whenever the operations involved can be performed.

- (a) A(BC) = (AB)C,
- (b) A(B+C) = AB + AC,
- (c) (A + B)C = AC + BC,
- (d) k(AB) = (kA)B = A(kB),
- (e)  $I_m A = A = A I_n$ if *A* is  $m \times n$ .

Proof: Omitted.

Similarly we have algebraic rules for the transpose.

**Theorem 3.4** Let *A* and *B* be matrices. The following identities hold whenever the operations involved can be performed.

(a) 
$$(A^{T})^{T} = A$$
,

(b) 
$$(A + B)^{T} = A^{T} + B^{T}$$
,

(c) 
$$(kA)^{T} = k(A^{T}),$$

(d) 
$$(AB)^{T} = B^{T}A^{T}$$
,

(e) 
$$(A^m)^T = (A^T)^m$$
 for all integers  $m \ge 0$ .

<sup>6</sup> i.e. even for square  $n \times n$  matrices Aand B (so that AB and BA are defined and are matrices of the same size), it does not follow that AB and BA are equal.

<sup>7</sup> Can you give an example where this happens?

Proof: Omitted.

Theorem 3.5

- (a) If A is a square matrix then  $A + A^{T}$  is a symmetric matrix,
- (b) For any matrix A,  $AA^{T}$  and  $A^{T}A$  are symmetric matrices.

Proof: Omitted.

#### 3.3: The inverse of a matrix

**Definition** If *A* is an  $n \times n$  matrix, the *inverse* of *A* is an  $n \times n$  matrix A' such that

$$AA' = I_n$$
, and  $A'A = I_n$ .

If A' exists we say A is *invertible*. If no inverse exists, then we say that A is not invertible.

Example Consider the matrices 
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 7 \end{bmatrix}$$
 and  $B = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ .

**Theorem** 3.6 If an  $n \times n$  matrix A is invertible then its inverse is unique.

Proof: Omitted.

**Notation** If A is invertible we write  $A^{-1}$  for its inverse. **Important Warning.** We are not allowed to write  $\frac{1}{A}$  for the inverse of A. Matrices are not numbers, we have defined a notion of multiplication but not of division.

**Theorem 3.7** If *A* is an invertible  $n \times n$  matrix then the system of linear equations given by Ax = b has the unique solution given by  $\mathbf{x} = A^{-1}\mathbf{b}$ .

Proof: Omitted.

**Theorem** 3.8 If

$$A = \left[ \begin{array}{cc} a & b \\ c & d \end{array} \right]$$

then *A* is invertible if  $ad - bc \neq 0$ , in which case

$$A^{-1} = \frac{1}{ad - bc} \left[ \begin{array}{cc} d & -b \\ -c & a \end{array} \right].$$

If ad - bc = 0 then A is not invertible.

Proof: Omitted.

For general  $n \times n$  matrices, there are algorithms (using row reduction) for finding the inverse, but not convenient general formula as we have in the  $2 \times 2$  case.

**Definition** For a  $2 \times 2$  matrix

$$A = \left[ \begin{array}{cc} a & b \\ c & d \end{array} \right],$$

we call ad - bc the determinant<sup>8</sup> of A, so that

$$det(A) = ad - bc$$
.

8 See section 4.2, where we will discuss the determinant of general square matrices

Example Find the inverses if they exist of  $A = \begin{bmatrix} 1 & 5 \\ 2 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 5 \\ 2 & 4 \end{bmatrix}$  $\begin{bmatrix} 1 & 5 \\ 2 & 10 \end{bmatrix}.$ 

Example Solve the system

$$x + 5y = 3, 2x + 4y = 1$$

using the inverse of the coefficient matrix.

# Theorem 3.9

(a) If A is an invertible matrix, then  $A^{-1}$  is invertible and

$$(A^{-1})^{-1} = A.$$

(b) If A is an invertible matrix and  $c \neq 0$  is a scalar then cA is invertible and

$$(cA)^{-1} = \frac{1}{c}A^{-1}$$

(c) If A and B are invertible matrices of the same size, then AB is invertible and

$$(AB)^{-1} = B^{-1}A^{-1}$$
.

(d) If A is an invertible matrix, then  $A^{T}$  is invertible and

$$(A^{\mathrm{T}})^{-1} = (A^{-1})^{\mathrm{T}}.$$

(e) If *A* is invertible matrix then  $A^n$  is invertible for all integers  $n \ge 0$ and

$$(A^n)^{-1} = (A^{-1})^n.$$

Proof: Omitted.

**Definition** If *A* is invertible and  $n \ge 0$  an integer we define

$$A^{-n} = (A^{-1})^n = (A^n)^{-1}.$$