

2.1 Matrix Operations

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Basic terminology

An $n \times m$ matrix A can be written as an $n \times m$ array of numbers.

It can also be written as

$$A = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_m]$$

where each \mathbf{a}_i is one of the m columns of A , and is a vector in \mathbf{R}^n .

Special matrices

The $m \times n$ zero matrix has all entries equal to zero.

The main diagonal of an $m \times n$ matrix are the entries a_{11}, a_{22}, \dots (which ends at either a_{nn} or a_{mm} depending on which is smaller).

A square matrix is diagonal if all entries off the main diagonal are zero.

Addition

Assuming matrices A and B **are the same shape** you can add them element by element:

$$C = A + B$$

and obtain another matrix of the same shape.

Scalar multiplication

You can also multiply A by a constant r (meaning multiply every element by r) to get another matrix $B = rA$ of the same shape.

Properties

Let A , B , and C be matrices of the same size, and let r and s be scalars.

a. $A + B = B + A$

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

c. $A + 0 = A$

d. $r(A + B) = rA + rB$

e. $(r + s)A = rA + sA$

f. $r(sA) = (rs)A$

Figure 1: properties

Matrix Multiplication

If A is an $m \times n$ matrix and B is a $k \times p$ matrix, then the product AB is defined ONLY WHEN $n = k$.

In other words you can multiply $m \times n$ times $n \times p$.

The result is an $m \times p$ matrix.

Matrix multiplication

If A is an $m \times n$ matrix and B is an $n \times p$ matrix:

$$B = [b_1 \quad b_2 \quad \cdots \quad b_p]$$

then

$$AB = [Ab_1 \quad Ab_2 \quad \cdots \quad Ab_p]$$

This makes sense because:

- ▶ each column b_1 has n rows
- ▶ Ab_1 makes sense and has m rows
- ▶ so AB is an $m \times p$ matrix.

Example

$$A = \begin{bmatrix} 8 & 6 & -8 & -4 \\ 0 & 4 & -4 & 8 \end{bmatrix}$$

$$B = \begin{bmatrix} 8 & 7 & 1 \\ 7 & 0 & 7 \\ 2 & 1 & -7 \\ -7 & 8 & -3 \end{bmatrix}$$

$$AB = \begin{bmatrix} 118 & 16 & 118 \\ -36 & 60 & 32 \end{bmatrix}$$

Dot product rule

If A is $m \times n$ and B is $n \times p$, then the i, j entry of AB is the dot product of row i of A with column j of B .

$$\text{Row}_i(A) = [a_{i1}, a_{i2}, \dots, a_{in}]$$

$$\text{Col}_j(B) = \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix}$$

$$A_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj}$$

Summation Notation

Here A is $m \times n$ and B is $n \times p$. $(AB)_{ij}$ means the i, j entry of the matrix product AB .

$$(AB)_{ij} = \sum_{t=1}^n a_{it} b_{tj}$$

Properties

Let A be an $m \times n$ matrix, and let B and C have sizes for which the indicated sums and products are defined.

- a. $A(BC) = (AB)C$ (associative law of multiplication)
- b. $A(B + C) = AB + AC$ (left distributive law)
- c. $(B + C)A = BA + CA$ (right distributive law)
- d. $r(AB) = (rA)B = A(rB)$
for any scalar r
- e. $I_m A = A = A I_n$ (identity for matrix multiplication)

Figure 2: MatrixMultProps

IMPORTANT: In general, $AB \neq BA$. Matrix multiplication is NOT COMMUTATIVE.

Transpose

The *transpose* of an $m \times n$ matrix A is the $n \times m$ matrix obtained from A by interchanging rows and columns.

$$A = \begin{bmatrix} -3 & -2 & -1 & 0 & 1 \\ 0 & -3 & 1 & -1 & -2 \\ -2 & -2 & -2 & -2 & -2 \end{bmatrix}$$

$$\text{Transpose } A^T = \begin{bmatrix} -3 & 0 & -2 \\ -2 & -3 & -2 \\ -1 & 1 & -2 \\ 0 & -1 & -2 \\ 1 & -2 & -2 \end{bmatrix}$$

Transpose and products

The transpose of a product is the product of the transposes, in the reversed order.

$$(AB)^T = B^T A^T$$

This makes sense: If A is $m \times n$ and B is $n \times p$, then (AB) is $m \times p$ so $(AB)^T$ is $p \times m$.

On the other hand B^T is $p \times n$ and A^T is $n \times m$ so $B^T A^T$ is also $p \times m$.