

Matrices and Vectors: Basic Introduction

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Overview

- 1 Definition of Matrices
- 2 Matrix Properties
- 3 Matrix Arithmetic
- 4 Definition of Vectors
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Definition of Matrices

Definition of a Matrix

Definition. A matrix (or array) of order m by n is simply a set of numbers arranged in a rectangular block of m horizontal rows and n vertical columns. We say

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

is a matrix of size (or dimension) $(m \times n)$.

In the double subscript notation a_{ij} for matrix element $a(i, j)$, the first subscript i denotes the row number, and the second subscript j denotes the column number.

Matrix Properties

Matrix Properties

Properties of Matrix A:

- A matrix having the same number of rows and columns is called **square**.
- A square matrix of order n is also called a $(n \times n)$ matrix.
- The elements $a_{11}, a_{22}, \dots, a_{nn}$ are called the **principal diagonal**.
- A diagonal matrix with elements $a_{ii} = 1$, and all other matrix elements zero, is called the identity matrix I .

Matrix Transpose

Matrix Transpose. The **transpose** of a $(m \times n)$ matrix A is the $(n \times m)$ matrix obtained by interchanging the rows and columns of A . The transpose is denoted A^T .

Example 1. The matrix transpose of

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \end{bmatrix} \quad \text{is} \quad A^T = \begin{bmatrix} 1 & 5 \\ 2 & 6 \\ 3 & 7 \\ 4 & 8 \end{bmatrix} \quad (2)$$

Properties

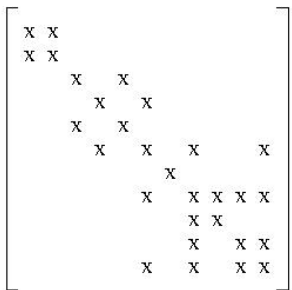
- $(A + B)^T = A^T + B^T$.
- $(ABC)^T = C^T B^T A^T$.

Symmetric and Skew-Symmetric Matrices

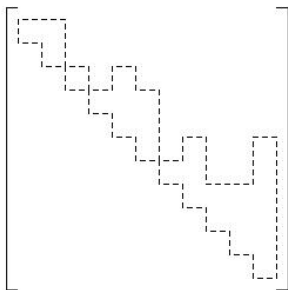
Matrix Symmetry:

- A square matrix A is **symmetric** if $A = A^T$.
- A square matrix A is **skew-symmetric** if $A = -A^T$.

Large symmetric matrices play a central role in structural analysis.



Schematic of Non-Zero Matrix Elements



Skyline Storage Pattern

Matrix Inverse

Definition: When it exists, the **inverse of matrix A** is written A^{-1} and it has the property:

$$[A] [A^{-1}] = [A^{-1}] [A] = I. \quad (3)$$

Nomenclature

- If matrix A has an inverse, then A is called **non-singular**.
- If matrix A has an inverse, then the inverse will be unique.
- If matrix A does not have an inverse, then A is called **singular**.

Theorem. For a $(n \times n)$ matrix A, the inverse A^{-1} exists \iff $\text{rank}(A) = n$.

- Conversely, matrix A is **singular** if $\text{rank}(A) < n$ (i.e., rank deficient).

Matrix Inverse

Computational Procedure. We want to carry out row operations such that:

$$[A|I] \xrightarrow{\text{row operations}} [I|A^{-1}]. \quad (4)$$

Example. Can apply row operations to get:

$$\left[\begin{array}{ccc|ccc} -1 & 1 & 2 & 1 & 0 & 0 \\ 3 & -1 & 1 & 0 & 1 & 0 \\ -1 & 3 & 4 & 0 & 0 & 1 \end{array} \right] \xrightarrow{\text{row ops}} \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -0.7 & 0.2 & 0.3 \\ 0 & 1 & 0 & -1.3 & -0.2 & 0.7 \\ 0 & 0 & 1 & 0.8 & 0.2 & -0.2 \end{array} \right]. \quad (5)$$

If A has $\text{rank}(A) < n$, then the last row in echelon form will be the O (zero) vector, and the **computation will fail**.

Matrix Inverse

Properties:

$$[A^{-1}]^{-1} = A. \quad (6)$$

$$(AB)^{-1} = B^{-1}A^{-1}. \quad (7)$$

$$(ABC)^{-1} = C^{-1}B^{-1}A^{-1}. \quad (8)$$

$$[A^T]^{-1} = [A^{-1}]^T. \quad (9)$$

Lower and Upper Triangular Matrices

A lower triangular matrix L is one where $a_{ij} = 0$ for all entries above the diagonal.

An upper triangular matrix U is one where $a_{ij} = 0$ for all entries below the diagonal. That is,

$$L = \begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \quad U = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{mn} \end{bmatrix} \quad (10)$$

Matrix Arithmetic

Matrix Addition and Subtraction

Definition. If A is a $(m \times n)$ matrix and B is a $(r \times p)$ matrix, then the matrix sum $C = A + B$ is defined only when $m = r$ and $n = p$, and is a $(m \times n)$ matrix C whose elements are

$$c_{ij} = a_{ij} + b_{ij}, \text{ for } i = 1, 2, \dots, m \text{ and } j = 1, 2, \dots, n. \quad (11)$$

Properties

- $(kA) B = k (A.B)$
- $A(BC) = (AB)C.$
- $(A + B)C = AB + AC.$
- $C(A + B) = CA + CB.$

Matrix Addition and Subtraction

Example 1. Let

$$A = \begin{bmatrix} 2 & 1 \\ 4 & 6 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 4 & 2 \\ 0 & 1 \end{bmatrix}. \quad (12)$$

The matrix sum is:

$$C = A + B = \begin{bmatrix} 2 & 1 \\ 4 & 6 \end{bmatrix} + \begin{bmatrix} 4 & 2 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix}. \quad (13)$$

Matrix Multiplication

Definition. Let A and B be $(m \times n)$ and $(r \times p)$ matrices, respectively.

The matrix product $A \cdot B$ is defined only when interior matrix dimensions are the same (i.e., $n = r$).

The matrix product $C = A \cdot B$ is a $(m \times p)$ matrix whose elements are

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj} \quad (14)$$

for $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$.

Matrix Multiplication

Example 1. Assuming that matrices A and B are as defined in the previous section:

$$\begin{aligned} C = A \cdot B &= \begin{bmatrix} 2 & 1 \\ 4 & 6 \end{bmatrix} \cdot \begin{bmatrix} 4 & 2 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 2 \cdot 4 + 1 \cdot 0 & 2 \cdot 2 + 1 \cdot 1 \\ 4 \cdot 4 + 6 \cdot 0 & 4 \cdot 2 + 6 \cdot 1 \end{bmatrix} && (15) \\ &= \begin{bmatrix} 8 & 5 \\ 16 & 14 \end{bmatrix}. \end{aligned}$$

Geometric Interpretation. Matrix element c_{ij} is the **dot product** of the **i-th row** of A with the **j-th column** of B.

Matrix Multiplication

Properties.

- $A.B.C = (A.B).C = A.(B.C).$
- $A.(B + C) = A.B + A.C.$
- $(A + B).C = A.C + B.C.$
- $A.I = A.$
- In general, $A.B \neq B.A.$
- $A.B = \phi$ does not necessarily imply $A = \phi$ or $B = \phi$. Counter example:

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} \text{ and } B = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}. \quad (16)$$

Definition of Vectors

Definition of Row and Column Vectors

Definition. A row vector is simply a $(1 \times n)$ matrix, i.e.,

$$V = [v_1 \quad v_2 \quad v_3 \quad v_4 \quad \cdots \quad v_n] \quad (17)$$

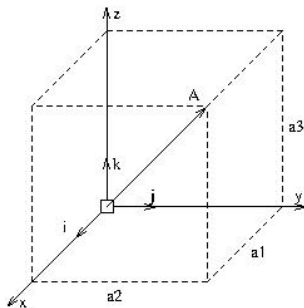
Definition. A column vector is a $(m \times 1)$ matrix, e.g.,

$$V = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ \cdots \\ v_m \end{bmatrix} \quad (18)$$

In both cases, the i -th element of the column vector is denoted v_i .

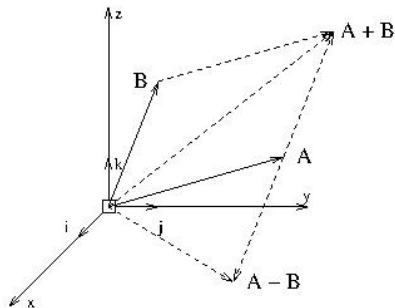
Vector Properties

Properties of Vector Arithmetic



Components of Three-Dimensional Vector

- $a + b = b + a$
- $a + 0 = a$
- $c(a + b) = ca + cb$



Vector Addition and Subtraction

- $(a + b) + c = a + (b + c)$
- $a + (-a) = 0$
- $1a = a$.

Dot Product

Definition. The dot product of two vectors $a = [a_1, a_2, a_3, \dots, a_n]$ and $b = [b_1, b_2, b_3, \dots, b_n]$ is:

$$a \cdot b = \sum_{i=1}^n a_i b_i = a_1 b_1 + a_2 b_2 + a_3 b_3 + \dots + a_n b_n. \quad (19)$$

Note: $a \cdot b = b \cdot a$. If a and b are perpendicular then $a \cdot b = 0$.

Engineering Applications

- Mechanical work is the dot product of force and displacement vectors (Jou).
- Power is the dot product of force and velocity vectors (W).
- Fluid Mechanics.

Dot Product

Example 1. Let $a = [1, 2, 3]$ and $b = [0, -1, 2]$. The dot product:

$$a \cdot b = \sum_{i=1}^n a_i b_i = 1 \times 0 + 2 \times -1 + 3 \times 2 = 4. \quad (20)$$

A dot product can also be written as a row vector multiplied by a column vector, e.g.,

$$[1, 2, 3] \cdot \begin{bmatrix} 0 \\ -1 \\ 2 \end{bmatrix} = 4. \quad (21)$$

The vector dimensions are: $(1 \times 3) (3 \times 1) \rightarrow (1 \times 1)$.

Dot Product

Properties. Let $a = [a_1, a_2, a_3, a_4]$, $b = [b_1, b_2, b_3, b_4]$ and $c = [c_1, c_2, c_3, c_4]$. And let d be a non-zero constant.

The dot product:

$$a \cdot b = a_1 b_1 + a_2 b_2 + a_3 b_3 + a_4 b_4 \quad (22)$$

obeys the properties:

- $a \cdot a = \|a\|^2$.
- $a \cdot (b + c) = a \cdot b + a \cdot c$
- $a \cdot b = b \cdot a$
- $a \cdot b = 0 \iff a = 0$ or $b = 0$ or $a \perp b$.
- $0 \cdot a = 0$
- $(da) \cdot b = d(a \cdot b)$
- $a \cdot b = |a| \cdot |b| \cos(\theta)$.

Cross Product

Definition. Consider two vectors A and B in three dimensions:

$$A = a_1\hat{i} + a_2\hat{j} + a_3\hat{k}$$

$$B = b_1\hat{i} + b_2\hat{j} + b_3\hat{k}$$

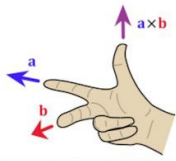
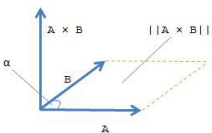
The cross product of A and B is:

$$\begin{aligned} C = A \times B &= \det \begin{pmatrix} \hat{i} & \hat{j} & \hat{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{pmatrix} \\ &= (a_2b_3 - a_3b_2)\hat{i} + (a_3b_1 - a_1b_3)\hat{j} + (a_1b_2 - a_2b_1)\hat{k}. \end{aligned}$$

Cross Product

Geometric Interpretation

$A \times B$ is a vector that is perpendicular to both A and B .



- The magnitude of $\|A \times B\|$ is equal to the area of the parallelogram formed using A and B as the sides.
- The angle between A and B is: $\|A \times B\| = \|A\| \|B\| \sin(\alpha)$.
- The cross product is zero when the A and B are parallel.

Linear Independence of Vectors

Linear Independence

A set of vectors $(v_1, v_2, v_3, \dots, v_n)$ is said to be **linearly independent** if the equation

$$a_1 v_1 + a_2 v_2 + a_3 v_3 + \dots + a_n v_n = 0. \quad (23)$$

can only be satisfied by $a_i = 0$ for $i = 1, \dots, n$.

Put another way: no vector in the sequence can be written as a linear combination of the other vectors.

Linear Independence of Vectors

Example 1. Consider three vectors $v_1 = (1, 1)$, $v_2 = (-3, 2)$, and $v_3 = (2, 4)$ in two-dimensional space.

The vectors will be **linearly independent** if the only solutions to

$$a_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + a_2 \begin{bmatrix} -3 \\ 2 \end{bmatrix} + a_3 \begin{bmatrix} 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}. \quad (24)$$

are $a_1 = a_2 = a_3 = 0$. Writing these equations in matrix form:

$$\begin{bmatrix} 1 & -3 & 2 \\ 1 & 2 & 4 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}. \quad (25)$$

Linear Independence of Vectors

Apply row operations (details to follow):

$$\begin{bmatrix} 1 & 0 & 16/5 \\ 0 & 1 & 2/5 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}. \quad (26)$$

which can be rearranged:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} + a_3 \begin{bmatrix} 16/5 \\ 2/5 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}. \quad (27)$$

We conclude that since a_1 and a_2 can be written in terms of a_3 , the equations are **linearly dependent**.

Linear Independence of Vectors

A Few Observations

- Vectors v_1 through v_3 are two dimensional.
- Can show that **three** (or more) **vectors** in **two-dimensional space** will always be **linearly dependent**.
- Can show that **four** (or more) **vectors** in **three-dimensional space** will always be **linearly dependent**.
- This is why a stool with three legs (**vectors**) will always be steady (**linearly independent**), but one with four legs (**vectors**) will sometimes rock (**linearly dependent**).

