CIS 678 Machine Learning

Introduction to Linear Algebra

Basic Math - Concept of Vectors, and Vector Space

We are aware of Scalars: A person's

Height (1.72m)

We are aware of Scalars: A person's

Height (1.72m) Weight (72kg)

We are aware of Scalars: A person's

Height (1.72m) Weight (72kg) Salary (100K)

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Height (1.72m) Weight (72kg) Salary (100K)

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A closed form definition of a person through some features

[Height (1.72m), Weight (72kg), Salary (100K)]

A closed form definition of a person through some features

- no explicit unit mentions

[1.72, 72, 100]

A closed form definition of a person through some features

- no explicit unit mentions

[1.72, 72, 100]

Is a vectoried representation of a person through some attributes: height, weight, salary

A closed form definition of a person through some features

- no explicit unit mentions

[1.72, 72, 100], [1.65, 70, 120], [1.81, 110, 90],

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[1.45, 65, 130],

And here we are talking about a number of people through same features: height, weight, salary

Formal definition of Vectors

Formal definition of Vectors

1. Vectors

We begin by defining a mathematical abstraction known as a **vector space**. In linear algebra the fundamental concepts relate to the *n***-tuples** and their algebraic properties.

Definition: An ordered *n*-tuple is considered as a sequence of *n* terms (a_1, a_2, \dots, a_n) , where *n* is a positive integer.

We see that an ordered *n*-tuple has terms whereas a set has members.

Example: A sequence (5) is called an ordered 1-tuple. A 2-tuple, for example (3, 6) (where 6 appears after 3) is called an ordered pair, and 3-tuple is called an ordered triple. A sequence (9, 3, 4, 4, 1) is called an ordered 5-tuple.

Let us denote the set of all ordered 1-tuples of real numbers by \mathbb{R} . We will write for example $(3.5) \in \mathbb{R}$.

X = [1.78, 72, 100]

$$\mathbf{x} = \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix} \qquad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

We are aware of Scalars: A person's height, weight, salary

Physics vector: velocity (scalar value + direction)

Algebraic vector (in general): Common representation of an entity (1 to n dimension):

- A person's (height, weight, salary), say [1.78, 72, 100]: once defined, we have to follow it.

$$\mathbf{x} = \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix} \qquad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

Vector operation rules

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1. \mathbf{x} + \mathbf{y} \in \mathbb{R}^{n}

2. \alpha \cdot \mathbf{x} \in \mathbb{R}^{n}

3. \mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x} \in \mathbb{R}^{n} (commutativity)

4. \alpha \cdot (\mathbf{x} + \mathbf{y}) = \alpha \cdot \mathbf{x} + \alpha \cdot \mathbf{y} (distributivity)

5. (\alpha + \beta) \cdot \mathbf{x} = \alpha \cdot \mathbf{x} + \beta \cdot \mathbf{x} (distributivity)

6. (\mathbf{x} + \mathbf{y}) + \mathbf{z} = \mathbf{x} + (\mathbf{y} + \mathbf{z}) (associativity)

7. (\alpha\beta) \cdot \mathbf{x} = \alpha \cdot (\beta \cdot \mathbf{x}) (associativity)
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Vector Operation

1.1.2. Vector Addition

Addition of vectors is defined:

$$\mathbf{x} + \mathbf{y} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_n + y_n \end{bmatrix}$$

Example:

$$\mathbf{x} + \mathbf{y} = \begin{bmatrix} 2 \\ 6 \\ -5 \end{bmatrix} + \begin{bmatrix} 0 \\ 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 2 \\ 9 \\ -1 \end{bmatrix}$$

Vector Operation

1.1.4. Zero Vector

The **zero** vector **sometimes denoted 0** is a vector having all elements equal to zero, e.g., the 2-dimensional **0** vector:

$$\mathbf{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \tag{A.7}$$

Vector Operation

1.1.9. Inner Product

The inner or dot product of two vectors x and y of the same dimension is a scalar defined by:

$$\mathbf{x}^T \cdot \mathbf{y} = (\mathbf{x}, \mathbf{y}) = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = \sum_{i=1}^n x_i y_i$$
 (A.11)

Note that the inner product of vector \mathbf{x} and \mathbf{y} requires that a transposed vector \mathbf{x} be multiplied by the \mathbf{y} vector. Sometimes the inner product is denoted simply by juxtaposition of the vectors x and y, for example, as $\langle \mathbf{x}, \mathbf{y} \rangle$ or (\mathbf{x}, \mathbf{y}) .

Example: The inner product of two vectors $\mathbf{x} = \begin{bmatrix} 4 \\ 1 \\ 7 \end{bmatrix}$ and $\mathbf{y} = \begin{bmatrix} 0 \\ 2 \\ -3 \end{bmatrix}$

$$\mathbf{x}^{T}\mathbf{y} = \begin{bmatrix} 4 \ 1 \ 7 \end{bmatrix}^{T} \begin{bmatrix} 0 \\ 2 \\ -3 \end{bmatrix} = 4 \cdot 0 + 1 \cdot 2 + 7 \cdot (-3) = 19$$

Vector Operation

1.1.10. Orthogonal Vectors

Two vectors \mathbf{x} and \mathbf{y} are said to be **orthogonal** if their inner product is equal to zero

$$\mathbf{x}^T \mathbf{y} = 0 \tag{A.12}$$

here 0 is a scalar.

Example: Two vectors $\mathbf{x} = \begin{bmatrix} 4 \\ 0 \end{bmatrix}$ and $\mathbf{y} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$ and are orthogonal, since their inner product is equal to zero

$$\mathbf{x}^T \cdot \mathbf{y} = \begin{bmatrix} 4 \\ 0 \end{bmatrix}^T = \begin{bmatrix} 0 \ 2 \end{bmatrix} = 4 \cdot 0 + 0 \cdot 2 = 0$$

Vector Operation

1.1.11. Vector Norm

The magnitude of a vector may be measure in different ways. One method, called the vector **norm**, is a function from \mathbb{R}^n into \mathbb{R} for \mathbf{x} an element of \mathbb{R}^n . It is denoted $||\mathbf{x}||$ and satisfies the following conditions:

- 1. $||\mathbf{x}|| \ge 0$, and the equality holds if and only if $\mathbf{x} = \mathbf{0}$
- 2. $||\alpha \mathbf{x}|| = |\alpha| \cdot ||\mathbf{x}||$, where $|\alpha|$ is the absolute value of scalar α

and is defined as:

$$||\mathbf{x}|| = \sqrt{\mathbf{x}^T \mathbf{x}} = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$
 (A.13)

Example: For the vector $\mathbf{x} = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$ the norm is

$$||\mathbf{x}|| = \sqrt{\mathbf{x}^T \mathbf{x}} = \sqrt{4^2 + 3^2} = 5$$

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