

# Euclidean Spaces

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## 1 Points and Vectors in Euclidean Space

### Objects in Euclidean Spaces

#### Objects in $n$ -dimensional Euclidean Spaces

Dimension	Object	Representation
0	point	$\emptyset$
1	line	$x_1 \in \mathbb{R}^1$
2	plane	$(x_1, x_2) \in \mathbb{R}^2$
3	3d space	$(x_1, x_2, x_3) \in \mathbb{R}^3$
$n$	$nd$ space	$(x_1, \dots, x_n) \in \mathbb{R}^n$

## 2 Vectors

### Vector

**Definition 1** ((Euclidean) Vector, displacement).  $n$ -tuples of real numbers  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  are (Euclidean) Vectors that represent displacement in  $\mathbb{R}^n$  space (or Cartesian coordinate system)

Let coordination of  $\mathbf{p} = (p_1, \dots, p_n)$ ,  $\mathbf{q} = (q_1, \dots, q_n)$ . Then the displacement from  $\mathbf{p}$  to  $\mathbf{q}$  is defined as  $\vec{\mathbf{pq}} := (q_1 - p_1, \dots, q_n - p_n)$ . In this definition,  $\mathbf{p}$  is an origin, and  $\mathbf{q}$  is a destination.

Note: Any vector  $\mathbf{p} = (p_1, \dots, p_n)$  can be interpreted as a location  $((p_1, \dots, p_n))$  or, displacement with origin  $\mathbf{0} := (0, \dots, 0)$  (more explicit notation:  $\begin{pmatrix} p_1 \\ \vdots \\ p_n \end{pmatrix}$ )

## 3 The Algebra of Vectors

### Addition and Subtraction

Let  $\mathbf{u}, \mathbf{v}$  be the vectors in  $\mathbb{R}^n$  space and  $u_i, v_i \in \mathbb{R}^1$  be their  $i$ -th element.

**Definition 2** ( $\pm$  of vectors).

$$(\mathbf{u} \pm \mathbf{v})_i := u_i \pm v_i \quad \forall i$$

or,

$$\mathbf{u} \pm \mathbf{v} := (u_1 \pm v_1, \dots, u_n \pm v_n)$$

Geometrically, addition of vectors means a sequence of displacements.

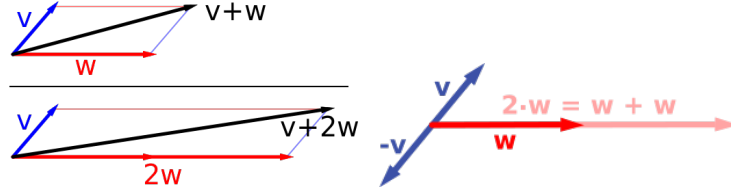


Figure 1: Geometrical Meaning: Addition, Subtraction, and Scalar Multiplication

### Scalar Multiplication

Let  $r, s$  be scalars (or real numbers). *i.e.*,  $r, s \in \mathbb{R}^1$ .

**Definition 3** (Scalar Multiplication).

$$(r\mathbf{u})_i := ru_i \quad \forall i$$

Geometrically, scalar multiplication means stretching or shrinking.

### Algebraic Properties of Vector Operation

$$\begin{aligned} \mathbf{u} + \mathbf{v} &= \mathbf{v} + \mathbf{u} && \text{(Commutative Law)} \\ (r + s)\mathbf{u} &= r\mathbf{u} + s\mathbf{u} && \text{(Distributive Law 1)} \\ r(\mathbf{u} + \mathbf{v}) &= r\mathbf{u} + r\mathbf{v} && \text{(Distributive Law 2)} \end{aligned}$$

In fact, any set of objects with a vector addition and scalar multiplication satisfying above laws is called vector space and vector is defined as the element of vector space. (Vector is defined by operations and their laws)

## 4 Length and Inner Product in $\mathbb{R}^n$

### Length and Direction

**Definition 4** (Length of Vector  $\|\vec{pq}\|$ ).

$$\|\vec{pq}\| := \sqrt{\sum_i^n (q_i - p_i)^2}$$

**Theorem 1** (10.1).

$$\|r\mathbf{v}\| = |r| \cdot \|\mathbf{v}\| \quad \forall r \in \mathbb{R} \wedge \forall \mathbf{v} \in \mathbb{R}^n$$

Any vector has two kinds of information: (1) length, and (2) direction.

**Definition 5** (Unit Vector (Direction of a vector)).

$$\text{Unit vector of } \mathbf{v} := \frac{\mathbf{v}}{\|\mathbf{v}\|}$$

## The Inner Product

**Definition 6** (Euclidean Inner Product (or dot product)).

$$\mathbf{u} \bullet \mathbf{v} := \sum_i^n u_i v_i \in \mathbb{R}^1$$

**Theorem 2** (10.2: Properties of Inner Product).    1.  $\mathbf{u} \bullet \mathbf{v} = \mathbf{v} \bullet \mathbf{u}$

2.  $\mathbf{u} \bullet (\mathbf{v} + \mathbf{w}) = \mathbf{u} \bullet \mathbf{v} + \mathbf{u} \bullet \mathbf{w}$

3.  $\mathbf{u} \bullet (r\mathbf{v}) = r(\mathbf{u} \bullet \mathbf{v}) = (r\mathbf{u}) \bullet \mathbf{v}$

4.  $\mathbf{u} \bullet \mathbf{u} \geq 0$

5.  $\mathbf{u} \bullet \mathbf{u} = 0 \iff \mathbf{u} = \mathbf{0}$

6.  $(\mathbf{u} + \mathbf{v}) \bullet (\mathbf{u} + \mathbf{v}) = \mathbf{u} \bullet \mathbf{u} + 2\mathbf{u} \bullet \mathbf{v} + \mathbf{v} \bullet \mathbf{v}$

## Inner Product and Angle between Two Vectors

**Theorem 3** (10.3). *Let  $\theta$  be the angle between  $\mathbf{u}, \mathbf{v}$ . Then,*

$$\mathbf{u} \bullet \mathbf{v} = \|\mathbf{u}\| \cdot \|\mathbf{v}\| \cos \theta$$

**Theorem 4** (10.4).    1.  $\theta$  is acute if  $\mathbf{u} \bullet \mathbf{v} > 0$

2.  $\theta$  is obtuse if  $\mathbf{u} \bullet \mathbf{v} < 0$

3.  $\theta$  is right if  $\mathbf{u} \bullet \mathbf{v} = 0$

**Theorem 5** (10.5: Triangle Inequality).

$$\|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\|$$

$$|\|\mathbf{u}\| - \|\mathbf{v}\|| \leq \|\mathbf{u} - \mathbf{v}\|$$

## Norms

**Definition 7** (Norms). *Any operation ( $X$ ) of a vector to a real number satisfying below three properties is norm. Length of vector is a norm.*

1.  $X(\mathbf{u}) \geq 0 \wedge X(\mathbf{u}) = 0$  only when  $\mathbf{u} = \mathbf{0}$

2.  $X(r\mathbf{u}) = |r|X(\mathbf{u})$

3.  $X(\mathbf{u} + \mathbf{v}) \leq X(\mathbf{u}) + X(\mathbf{v})$

Norm is set of distance measures between two vectors.

## 5 Lines and Planes

### 1D,2D Objects in $\mathbb{R}^n$ Spaces

#### Lines: One dimensional objects in $\mathbb{R}^n$ Spaces

$\mathbf{x}$  representing a line passing  $\overline{\mathbf{x}_0}$  with direction  $\overline{\mathbf{v}}$  is:

$$\mathbf{x} = \overline{\mathbf{x}_0} + t\overline{\mathbf{v}} \quad \forall t \in \mathbb{R} \quad (\text{Parametric Representation})$$

$\mathbf{x}$  representing a line passing  $\overline{\mathbf{x}_0}, \overline{\mathbf{x}_1}$  is:

$$\mathbf{x} = (1-t)\overline{\mathbf{x}_0} + t\overline{\mathbf{x}_1} \quad \forall t \in \mathbb{R} \quad (\text{Parametric Representation})$$

#### Planes: Two dimensional objects in $\mathbb{R}^n$ Spaces

$\mathbf{x}$  representing a plane passing  $\overline{\mathbf{x}_0}$  with direction  $\overline{\mathbf{v}_1}$  and  $\overline{\mathbf{v}_2}$  is:

$$\mathbf{x} = \overline{\mathbf{x}_0} + t_1\overline{\mathbf{v}_1} + t_2\overline{\mathbf{v}_2} \quad \forall t_i \in \mathbb{R} \quad (\text{Parametric Representation})$$

$\mathbf{x}$  representing a plane containing  $\overline{\mathbf{x}_0}, \overline{\mathbf{x}_1}, \overline{\mathbf{x}_2}$  is:

$$\mathbf{x} = (1-t_1-t_2)\overline{\mathbf{x}_0} + t_1\overline{\mathbf{x}_1} + t_2\overline{\mathbf{x}_2} \quad \forall t_i \in \mathbb{R} \quad (\text{Parametric Representation})$$

### Nonparametric Equations

**Definition 8** (Normal Vector). A *normal vector*  $\overline{\mathbf{n}}$  of a  $n$ -dimensional object  $\mathbf{X}$  is a vector which is perpendicular to any vectors in the object. Suppose  $\mathbf{x}, \mathbf{p}$  are location vectors in the object. Then,

$$\overline{\mathbf{n}} \bullet (\mathbf{x} - \mathbf{p}) = 0 \quad \forall \mathbf{x}, \mathbf{p} \in \mathbf{X}$$

### Finding Normal Vector

1. For linearly independent vectors  $\mathbf{u}_i$ , solve below systems of equations satisfying

$$\overline{\mathbf{n}} \bullet \mathbf{u}_i = 0 \quad \forall i$$

2.  $\overline{\mathbf{n}} = \mathbf{u} \times \mathbf{v}$  (Only for  $\mathbb{R}^3$  space)

$$\mathbf{u} \times \mathbf{v} := \left( \begin{vmatrix} u_2 & u_3 \\ v_2 & v_3 \end{vmatrix}, -\begin{vmatrix} u_1 & u_3 \\ v_1 & v_3 \end{vmatrix}, \begin{vmatrix} u_1 & u_2 \\ v_1 & v_2 \end{vmatrix} \right)$$

### Hyperplanes

**Definition 9** (Hyperplane).  $n-1$  dimensional object in  $\mathbb{R}^n$  space is a hyperplane with normal vector  $\overline{\mathbf{a}}$  if:

$$\overline{a}_1x_1 + \overline{a}_2x_2 + \cdots + \overline{a}_nx_n = \overline{d}$$

or,

$$\overline{\mathbf{a}} \bullet \mathbf{x} = \overline{d}$$

## 6 Economic Applications

### Budget Constraint

**Definition 10** (Commodity Bundle, Price Vector, Budget Set). *Let  $x_i \geq 0$  be the quantity of  $i$ th commodity.*

$$\mathbf{x} := (x_1, \dots, x_n), \quad x_i \geq 0 \quad \forall i \quad (\text{Commodity Bundle})$$

*Let  $p_i \geq 0$  be the price of  $i$ th commodity.*

$$\mathbf{p} := (p_1, \dots, p_n), \quad p_i \geq 0 \quad \forall i \quad (\text{Price Vector})$$

*Then budget set  $\mathbf{x}$  can be defined with given budget  $\bar{I}$ :*

$$\bar{\mathbf{p}} \bullet \mathbf{x} \leq \bar{I} \quad (\text{Budget Constraint})$$