

Polyhedral geometry 2

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Section 1

Affine spaces

Affine space

The idea of affine space corresponds to that of a set of points where the **displacement** from a point \mathbf{x} to another point \mathbf{y} is obtained by summing a vector \mathbf{v} to the \mathbf{x} point.

Definition

A set \mathcal{A} of points is called an **affine space** modeled on the vector space \mathcal{V} if there is a function

$$\mathcal{A} \times \mathcal{V} \rightarrow \mathcal{A} : (\mathbf{x}, \mathbf{v}) \mapsto \mathbf{x} + \mathbf{v}$$

called **affine action**, with the properties:

1. $(\mathbf{x} + \mathbf{v}) + \mathbf{w} = \mathbf{x} + (\mathbf{v} + \mathbf{w})$ for each $\mathbf{x} \in \mathcal{A}$ and each $\mathbf{v}, \mathbf{w} \in \mathcal{V}$;

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2. $\mathbf{x} + \mathbf{0} = \mathbf{x}$ for each $\mathbf{x} \in \mathcal{A}$, where $\mathbf{0} \in \mathcal{V}$ is the null vector;
3. for each pair $\mathbf{x}, \mathbf{y} \in \mathcal{A}$ there is a unique $(\mathbf{y} - \mathbf{x}) \in \mathcal{V}$ such that

$$\mathbf{x} + (\mathbf{y} - \mathbf{x}) = \mathbf{y}.$$

Dimension

The affine space \mathcal{A} is said of **dimension** n if modeled on a vector space \mathcal{V} of dimension n .

Vector sum vs affine action

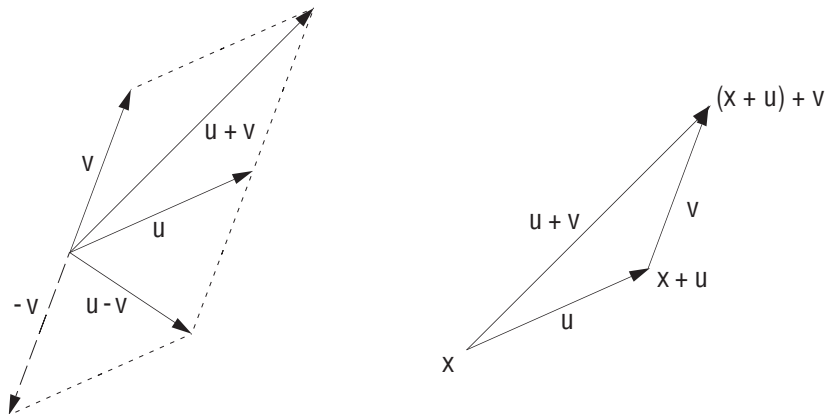


Figure: (a) Vector sum and difference are given by the parallelogram rule
(b) associativity of displacement (point and vector sum) in an affine space

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- ▶ Addition and difference of vectors are geometrically produced by the **parallelogram rule**
- ▶ notice also the **associative property** of the affine action on a point space.

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The sum of a set $\{\mathbf{v}_i\}$ of vectors ($i = 1, \dots, n$) can be geometrically obtained, in an affine space:

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Remark

1. the addition of points is not defined;
2. the difference of two points is a vector;
3. the sum of a point and a vector is a point.

Positive, affine and convex combinations

Three types of combinations of vectors or points can be defined. They lead to the concepts of **cones**, **hyperplanes** and **convex sets**, respectively.

Positive combination

Let $\mathbf{v}_0, \dots, \mathbf{v}_d \in \mathbb{R}^n$ and $\alpha_0, \dots, \alpha_d \in \mathbb{R}^+ \cup \{0\}$.

The vector

$$\alpha_0 \mathbf{v}_0 + \dots + \alpha_d \mathbf{v}_d = \sum_{i=0}^d \alpha_i \mathbf{v}_i$$

is called a **positive combination** of such vectors.

The set of all the positive combinations of $\{\mathbf{v}_0, \dots, \mathbf{v}_d\}$ is called the **positive hull** of $\{\mathbf{v}_0, \dots, \mathbf{v}_d\}$ and denoted $\text{pos}\{\mathbf{v}_0, \dots, \mathbf{v}_d\}$.

This set is also called the **cone** generated by the given vectors

Affine combination

Let $\mathbf{p}_0, \dots, \mathbf{p}_d \in \mathbb{E}^n$ and $\alpha_0, \dots, \alpha_d \in \mathbb{R}$, such that $\alpha_0 + \dots + \alpha_d = 1$.

The point

$$\sum_{i=0}^d \alpha_i \mathbf{p}_i := \mathbf{p}_0 + \sum_{i=1}^d \alpha_i (\mathbf{p}_i - \mathbf{p}_0)$$

is called an **affine combination** of the points $\mathbf{p}_0, \dots, \mathbf{p}_d$.

Affine combination

The set of all affine combinations of $\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$ is an **affine subspace**, denoted by $\text{aff}\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$

It is easy to verify that:

$$\text{aff}\{\mathbf{p}_0, \dots, \mathbf{p}_d\} = \mathbf{p}_0 + \text{lin}\{\mathbf{p}_1 - \mathbf{p}_0, \dots, \mathbf{p}_d - \mathbf{p}_0\}.$$

Affine combination

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Double description

Every affine subspace can be described either as

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Double description

Every affine subspace can be described either as

- ▶ the **intersection** of affine **hyperplanes**, or as
- ▶ the **affine hull** of a finite set of **points**.

Convex combination

Let $\mathbf{p}_0, \dots, \mathbf{p}_d \in \mathbb{E}^n$ and $\alpha_0, \dots, \alpha_d \geq 0$, with $\alpha_0 + \dots + \alpha_d = 1$.

The point

$$\alpha_0 \mathbf{p}_0 + \dots + \alpha_d \mathbf{p}_d = \sum_{i=0}^d \alpha_i \mathbf{p}_i$$

is called a **convex combination** of points $\mathbf{p}_0, \dots, \mathbf{p}_d$.

A **convex** combinations is both **positive** and **affine**.

Convex hull

The set of **all** convex combinations of $\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$ is a convex set, called **convex hull** of $\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$, and is denoted by $\text{conv}\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$.

Properties

- ▶ the convex hull of a set of points is the **intersection of all convex sets** that contain them

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Properties

- ▶ the convex hull of a set of points is the **intersection of all convex sets** that contain them
- ▶ the convex hull of a set of points is the **smallest set** that contains them

Computational Geometry

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- ▶ since very large datasets contain tens or hundreds of millions of points, the focus is on **computational complexity**
- ▶ for large data sets, the difference between $O(n^2)$ and $O(n \log n)$ may be between **days** and **seconds** of computation.

Convex hull algorithms: Gift wrapping (or Jarvis march)

$O(nh)$: where n is the number of points and h is the number of points on the convex hull

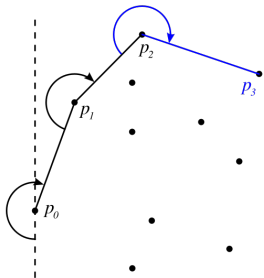


Figure: begin with $i = 0$, at point p_0 known to be on the convex hull, and select the point p_{i+1} such that all points are to the right of the line $[p_i, p_{i+1}]$

Linear, affine and convex independence

A set of **vectors** is said to be **linearly independent** if none of them can be obtained as a linear combination of the other ones.

Analogously, a set of **points** is said **affinely/convexly independent** if none of them can be obtained as an affine/convex combination of the other ones, respectively.

Affine combination

The set of **affine combinations** of **two affinely independent** (i.e. **distinct**) **points** $\mathbf{p}_0, \mathbf{p}_1 \in \mathbb{E}^n$, $\mathbf{p}_0 \neq \mathbf{p}_1$, is **their line**:

$$\begin{aligned}\alpha_0 \mathbf{p}_0 + \alpha_1 \mathbf{p}_1 &:= (1 - \alpha_1) \mathbf{p}_0 + \alpha_1 \mathbf{p}_1 \\ &= \mathbf{p}_0 + \alpha_1 (\mathbf{p}_1 - \mathbf{p}_0)\end{aligned}$$

Remember that a difference of points is a vector, and that the sum of a point and a vector is a point.

Convex combination (examples)

1. The set of **convex combinations** of **two affinely independent (i.e. distinct) points** $\mathbf{p}_0, \mathbf{p}_1 \in \mathbb{E}^n$, $\mathbf{p}_0 \neq \mathbf{p}_1$, is the **line segment** joining \mathbf{p}_0 with \mathbf{p}_1 :

$$\begin{aligned}\mathbf{p}(\beta) &= (1 - \beta)\mathbf{p}_0 + \beta\mathbf{p}_1 \\ &= \mathbf{p}_0 + \beta(\mathbf{p}_1 - \mathbf{p}_0), \quad 0 \leq \beta \leq 1\end{aligned}$$

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1. The set of **affine combinations** of **three affinely independent (i.e. not aligned) points** is their **plane**.
2. The set of **convex combinations** of **three affinely independent points** is their triangle, i.e. the **triangle** whose vertices are those points.

Plasm.js: Exercise 10 (Plane for three points)

Let $\mathbf{p}_0, \mathbf{p}_1, \mathbf{p}_2 \in \mathbb{E}^3$ be **three non collinear points**, i.e. such that:

$$\det \begin{pmatrix} \mathbf{p}_0 & \mathbf{p}_1 & \mathbf{p}_2 \end{pmatrix} \neq 0.$$

The vector \mathbf{n} , normal to the bundle of planes parallel to the 3 points, may be computed as

$$\mathbf{n} = (\mathbf{p}_1 - \mathbf{p}_0) \times (\mathbf{p}_2 - \mathbf{p}_0)$$

The Cartesian equation $ax + by + cz + d = 0$ of the plane for $\mathbf{p}_0, \mathbf{p}_1, \mathbf{p}_2$, with

$$a = n_x, \quad b = n_y, \quad c = n_z$$

is finally fixed by imposing the passage for one of such points—say $\mathbf{p}_0 = (x_0, y_0, z_0)^T$:

$$ax_0 + by_0 + cz_0 + d = 0 \quad \text{and hence} \quad d = -(n_x x_0 + n_y y_0 + n_z z_0)$$

Affine coordinates

Given a point $\mathbf{p} \in \text{aff}\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$, the scalars $\alpha_0, \dots, \alpha_d$ such that:

$$\mathbf{p} = \alpha_0 \mathbf{p}_0 + \dots + \alpha_d \mathbf{p}_d,$$

with $\alpha_0 + \dots + \alpha_d = 1$, are called **affine coordinates** of \mathbf{p} .

Affine coordinates are **unique** if the points $\mathbf{p}_0, \dots, \mathbf{p}_d$ are affinely independent.

Remarks

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2. Points of the d -flat are associated bijectively with such $(d + 1)$ -tuples of coordinates

Convex coordinates

Given a point $\mathbf{p} \in \text{conv}\{\mathbf{p}_0, \dots, \mathbf{p}_d\}$, the scalars $\alpha_0, \dots, \alpha_d$ such that:

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with $\alpha_0, \dots, \alpha_d \geq 0$ and $\alpha_0 + \dots + \alpha_d = 1$, are called **convex coordinates** of \mathbf{p} .

Convex coordinates are **unique** if the points $\mathbf{p}_0, \dots, \mathbf{p}_d$ are affinely independent.

Affine coordinates

The **affine coordinates** of \mathbb{E}^d are $(d + 1)$ -tuples of numbers summing to one.

A linear mapping between \mathbb{E}^d points and the \mathbb{E}^{d+1} hyperplane $x_1 + x_2 + \dots + x_{d+1} = 0$ is used.

The **transformation of coordinates** depends on $d + 1$ affinely independent points $(\mathbf{p}_0, \dots, \mathbf{p}_d) \in \mathbb{E}^d$ that, embedded in $x_d = 1$, are mapped to the standard basis:

$$[\mathbf{I}] = [\mathbf{T}] \begin{pmatrix} \mathbf{p}_0 & \dots & \mathbf{p}_d \\ 1 & \dots & 1 \end{pmatrix}$$

so that

$$[\mathbf{T}] = \begin{pmatrix} \mathbf{p}_0 & \dots & \mathbf{p}_d \\ 1 & \dots & 1 \end{pmatrix}^{-1}$$