

# Mathematics Triple Record - Academic Section

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## 1 Functions, Combinations, Hulls, and Dependence

The terms linear, affine, and convex can apply to all the terms in the title of this section.

### 1.1 Linear

Examples of linear functions are  $f(x) = x$  and  $f(x, y, z) = 2x + y - z$ . There are two properties that they must fulfill:  $f(ax) = af(x)$ , and  $f(x+y) = f(x) + f(y)$ . It follows from the second property that  $f(\sum x_i) = \sum f(x_i)$ . Therefore, linear functions are of the form  $f(x_1, \dots, x_n) = \sum_{i=1}^n a_i x_i$ . A linear combination of a set of  $x_i$ s is  $\sum_{i=1}^n a_i x_i$ . A linear hull of a set of  $x_i$ s is the set of all possible linear combinations of the  $x_i$ s. The linear hull of 1 point is a line connecting the point and the origin, of 2 points is (in general) the plane connecting the points and the origin, etc. We have four different definitions for linear dependence:

1. Some  $x_i$  is a linear combination of the other  $x_j$ s.
2. Some point can be written as a linear combination of all of the  $x_i$ s in 2 different ways.
3.  $\exists \lambda_i \mid \sum_{i=1}^k \lambda_i x_i = 0$  with  $\lambda_i \neq 0$  for some  $i$ .
4.  $\exists A, B \mid A \subseteq S, B \subseteq S, A \neq B, \text{lin}(A) = \text{lin}(B)$

We have proven that all of these definitions are equivalent.

### 1.2 Affine

Examples of affine functions are  $f(x) = 2x + 3$  and  $f(x, y, z) = 3x - 4y + \frac{z}{2} + 5$ . They are linear functions with a constant term added. An affine combination of a set of  $x_i$ s is  $\sum_{i=1}^n \lambda_i x_i$  such that  $\sum_{i=1}^n \lambda_i = 1$ . An affine hull of a set of  $x_i$ s is the set of all possible affine combinations of the  $x_i$ s. The affine hull of 1 point is the point itself, of 2 points is a line connecting the points, of 3 points (in general) is the plane, etc. We have four different definitions for affine dependence:

1. Some  $x_i$  is an affine combination of the other  $x_j$ s.
2. Some point can be written as an affine combination of all of the  $x_i$ s in 2 different ways.

3.  $\exists \lambda_i \mid \sum_{i=1}^k \lambda_i x_i = 0$  with  $\sum_{i=1}^k \lambda_i = 0$  and  $\lambda_i \neq 0$  for some  $i$ .
4.  $\exists A, B \mid A \subseteq S, B \subseteq S, A \neq B, \text{aff}(A) = \text{aff}(B)$ .

We have proven that all of these definitions are equivalent.

### 1.3 Convex

Convex combinations are affine combinations such that each  $\lambda_i \in [0, 1]$ . A set is convex if all segments with both endpoints in the set are completely contained in the set. The convex hull of a set of  $x_i$ s is the set of all possible convex combinations of the  $x_i$ s. The convex hull of two points is the segment connecting the points (without extending beyond), of three points (in general) is the triangle with those points as vertices, etc.

## 2 Boundary / Interior

### 2.0.1 Boundary

A boundary is a  $(d - 1)$ -surface of an object in  $\mathbb{R}^d$ . Formally, a boundary is the intersection of the closure of a set and the closure of its complement.

### 2.0.2 Interior:

An interior is the the  $d$ -space in an object in  $\mathbb{R}^d$ . It is all points in a set that are not on the boundary of the set.

### 2.1 Relatives

#### 2.1.1 Relative Boundary

The relative boundary of an  $\ell$ -object in  $d$ -space, where  $\ell \leq d$ , is the absolute boundary of the object in  $\ell$ -space.

#### 2.1.2 Relative Interior

The relative interior of an  $\ell$ -object in a  $d$ -space, where  $\ell \leq d$ , is the absolute interior of the object in  $\ell$ -space.

### 2.2 Dot Product

A.K.A. *Standard Inner Product* or *Scalar Product*:

$$\langle \ , \ \rangle : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$$

$$\langle x, y \rangle = \sum_{i=1}^d x_i y_i$$

where  $x = (x_1, x_2, \dots, x_d)$  and  $y = (y_1, y_2, \dots, y_d)$ .

### 3 Polytopes

A polytope is defined as the convex hull of a finite number of points. Polytopes are classified as  $d$ -polytopes, where  $d$  denotes the number of dimensions of the polytope, or the  $\mathbb{R}^d$ -space that the polytope exists in. All  $\ell$ -dimensional substructures of a  $d$ -dimensional polytope should also be polytopes. Known polytopes are:

- $d$ -cubes: These are formed by “dragging out” a  $d - 1$ -cube, dragging the same distance through a new dimension every time.
- $d$ -simplices: These are formed by “dragging out” a  $d - 1$ -simplex the same distance through a new dimension, and converging to a point.
- $d$ -prisms: These are formed by dragging out a  $d - 1$ -dimensional polytope a distance  $I$ .
- $d$ -cross polytopes: These are formed by finding the convex hull of  $d$  copies of  $[-1,1]$  arranged orthogonally and meeting at the origin.
- $d$ -fold prisms: These are of the form  $(P_{d-n} \times \underbrace{I \times \dots \times I}_{n \text{ of these}})$
- $d$ -fold pyramids: These are of the form  $\text{conv}(P_{d-n}, \underbrace{*, \dots, *}_{n \text{ of these}})$ , where the  $n$  points are affinely independent of  $P_{d-n}$  and of each other.

### 4 Other Figures

Flat: A flat is a space. The progression of flats is a point, line, plane, space, etc. We have two potential formal definitions. The first, Michael’s, states that a  $k$ -flat is the linear hull of  $k$  linearly independent points in  $\mathbb{R}^d$ , with  $k \leq d$ . The other, Adam’s states that a  $k$ -flat is the affine hull of  $k + 1$  affinely independent points in  $\mathbb{R}^d$ , with  $k \leq d$ .

Hyperplane: A hyperplane is a  $(d - 1)$ -flat in  $\mathbb{R}^d$ , that is, a flat one dimension lower than the space in which it is contained. A supporting hyperplane is a hyperplane that intersects the relative boundary, but not the relative interior, of a convex set.

Halfspace: A halfspace is a hyperplane and all of the space to one side of it. It is defined as  $K(a, c) = \{x \mid \langle a, x \rangle \leq c\}$ . To get the other half of space, make  $a$  and  $c$  negative. When this is reevaluated, it comes out to  $K(a, c) = \{x \mid \langle a, x \rangle \geq c\}$ .