

Convex Sets and Convex Functions (part I)

Prof. Dan A. Simovici

UMB

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Special Subsets in \mathbb{R}^n

Let L be a real linear space and let $x, y \in L$. The **closed segment** determined by x and y is the set

$$[x, y] = \{(1 - a)x + ay \mid 0 \leq a \leq 1\}.$$

The **half-closed segments** determined by x and y are the sets

$$[x, y) = \{(1 - a)x + ay \mid 0 \leq a < 1\},$$

and

$$(x, y] = \{(1 - a)x + ay \mid 0 < a \leq 1\}.$$

The **open segment** determined by x and y is

$$(x, y) = \{(1 - a)x + ay \mid 0 < a < 1\}.$$

The **line** determined by x and y is the set

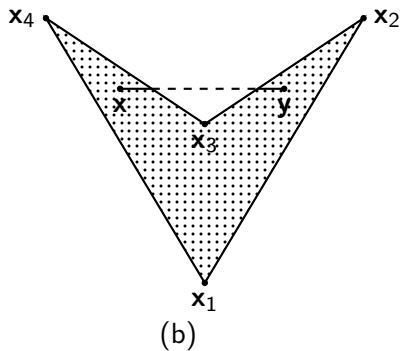
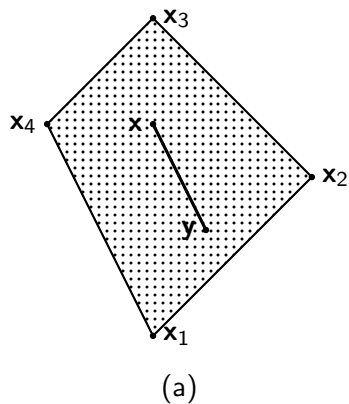
$$\ell_{x,y} = \{(1 - a)x + ay \mid a \in \mathbb{R}\}.$$

Definition

A subset C of L is **convex** if we have $[x, y] \subseteq C$ for all $x, y \in C$.

Note that the empty subset and every singleton $\{x\}$ of L are convex.

Convex vs. Non-convex



Example

The set $\mathbb{R}_{\geq 0}^n$ of all vectors of \mathbb{R}^n having non-negative components is a convex set called the **non-negative orthant** of \mathbb{R}^n .

Example

The convex subsets of $(\mathbb{R}, +, \cdot)$ are the intervals of \mathbb{R} . Regular polygons are convex subsets of \mathbb{R}^2 .

Example

Every linear subspace T of a real linear space L is convex.

Example

Let $(L, \|\cdot\|)$ be a normed linear space. An open sphere $B(x_0, r) \subseteq L$ is convex.

Indeed, suppose that $x, y \in B(x_0, r)$, that is, $\|x - x_0\| < r$ and $\|x_0 - y\| < r$.

Let $a \in [0, 1]$ and let $z = (1 - a)x + ay$. We have

$$\begin{aligned} \|x_0 - z\| &= \|x_0 - (1 - a)x - ay\| \\ &= \|a(x_0 - y) + (1 - a)(x_0 - x)\| \\ &\leq a\|x_0 - y\| + (1 - a)\|x_0 - x\| < r. \end{aligned}$$

so $z \in B(x_0, r)$.

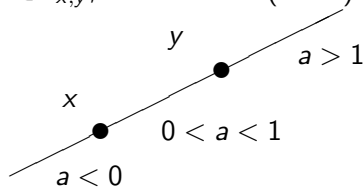
Similarly, a closed sphere $B[x_0, r]$ is a convex set.

Theorem

Let x, y, z be three distinct points in the real linear space L such that $z \in \ell_{x,y}$. Then, one of these points belongs to the open segment determined by the remaining two points.

Proof

Since $z \in \ell_{x,y}$, we have $z = (1 - a)x + ay$ for some $a \in \mathbb{R}$.



We have $a \notin \{0, 1\}$ because the points x, y, z are distinct. If $a > 1$ we have $y = \frac{a-1}{a}x + \frac{1}{a}z$, so $y \in (x, z)$ because $\frac{a-1}{a}, \frac{1}{a} \in (0, 1)$. If $0 < a < 1$ we have $z \in (x, y)$. Finally, if $a < 0$, since $x = \left(1 + \frac{a}{1-a}\right)z + \frac{-a}{1-a}y$, we have $x \in (z, y)$.

Definition

Let U be a subset of a real linear space L and let $x_1, \dots, x_k \in U$. A linear combination of U , $a_1x_1 + \dots + a_kx_k$, where $a_1, \dots, a_k \in \mathbb{R}$ and $k \geq 1$ is:

- an **affine combination** of U if $\sum_{i=1}^k a_i = 1$;
- a **non-negative combination** of U if $a_i \geq 0$ for $1 \leq i \leq k$;
- a **positive combination** of U if $a_i > 0$ for $1 \leq i \leq k$;
- a **convex combination** of U if it is both a non-negative and an affine combination of U .

Theorem

Let L be a real linear space. A subset C of L is convex if and only if any convex combination of elements of C belongs to C .

Proof

The sufficiency of this condition is immediate. To prove its necessity consider $x_1, \dots, x_k \in C$ and the convex combination

$$y = a_1x_1 + \cdots + a_kx_k.$$

We prove by induction on $k \geq 1$ that $y \in C$. The base case, $m = 1$ is immediate since in this case $y = a_1x_1$ and $a_1 = 1$.

For the inductive step, suppose that the statement holds for k and let y be given by $y = a_1x_1 + \cdots + a_kx_k + a_{k+1}x_{k+1}$, where $a_1 + \cdots + a_k + a_{k+1} = 1$, $a_i \geq 0$ and $x_i \in C$ for $1 \leq i \leq k+1$. We have

$$y = (1 - a_{k+1}) \sum_{i=1}^k \frac{a_i}{1 - a_{k+1}} x_i + a_{k+1}x_{k+1}.$$

Since $z = \sum_{i=1}^k \frac{a_i}{1 - a_{k+1}} x_i$ is a convex combination of k vectors, we have $z \in C$ by the inductive hypothesis, and the equality $y = (1 - a_{k+1})z + a_{k+1}x_{k+1}$ implies $y \in C$.

Definition

Let L, K be two linear spaces. A mapping $f : L \longrightarrow K$ is **affine** when there exists a linear mapping $h : L \longrightarrow K$ and some $b \in K$ such that $f(x) = h(x) + b$ for every $x \in L$

Theorem

Let L, K be two linear spaces and let $f : L \longrightarrow K$ be an affine mapping. If C be a convex subset of L , then $f(C)$ is a convex subset of K . If D is a convex subset of K , then $f^{-1}(D) = \{x \in L \mid f(x) \in D\}$ is a convex subset of L .

Proof

Since f is an affine mapping, we have $f(x) = h(x) + b$, where $h : L \rightarrow K$ is a linear mapping and $b \in K$ for $x \in L$. Therefore, if $y_1, y_2 \in f(C)$ we can write $y_1 = h(x_1) + b$ and $y_2 = h(x_2) + b$. This, in turn, allows us to write for $a \in [0, 1]$:

$$\begin{aligned}(1-a)y_1 + ay_2 &= (1-a)h(x_1) + (1-a)b + ah(x_2) + ab \\ &= h((1-a)x_1 + ax_2) + b = f((1-a)x_1 + ax_2).\end{aligned}$$

The convexity of C implies $(1-a)x_1 + ax_2 \in C$, so $(1-a)y_1 + ay_2 \in f(C)$, which shows that $f(C)$ is convex.

Definition

A subset C of a linear space L is **affine subspace** if $\ell_{x,y} \subseteq C$ for all $x, y \in C$.

In other words, C is a non-empty affine subspace if every point on the line determined by two members of C , x and y belongs to C . Note that C is a subspace of L if and only if $0_L \in C$ and C is an affine subspace.

Example

The empty set \emptyset , every singleton $\{x\}$, and the entire space L are affine subspaces of L . Also, every hyperplane H is an affine subspace of L .

Theorem

A non-empty subset C of a linear space L is an affine subspace if and only if any affine combination of elements of C belongs to C .

It is immediate to verify that any translation of a linear space K is an affine subspace. The converse is also true as we show next.

Theorem

Let D be a non-empty affine subspace in a linear space L . There exists a translation t_u and a unique subspace K of L such that $D = t_u(K)$.

Proof

Let $K = \{x - y \mid x, y \in D\}$ and let $x_0 \in D$. We have $0_L = x_0 - x_0 \in K$ and it is immediate that K is subspace of L .

Let u be an element of D . We claim that $D = t_u(K)$. Indeed, if $z \in D$, $z - u \in K$, so $z \in t_u(K)$, which implies $D \subseteq t_u(K)$. Conversely, if $x \in t_u(K)$ we have $x = u + v$ for some $v \in K$ and, therefore, $x = u + s - t$ for some $s, t \in D$, where $v = s - t$. This implies $x \in D$ because $u + s - t$ is an affine combination of D .

Proof (cont'd)

To prove the uniqueness of the subspace K suppose that $D = t_u(K_1) = t_v(K_2)$, where both K_1 and K_2 are subspaces of L . Since $0_L \in K_2$, it follows that there exists $w \in K_1$ such that $u + w = v$. Similarly, since $0_L \in K_1$, it follows that there exists $t \in K_1$ such that $u = v + t$, which implies $w + t = 0_L$. Thus, both w and t belong to both subspaces K_1 and K_2 . If $s \in K_1$, it follows that $u + s = v + z$ for some $z \in K_2$. Therefore, $s = (v - u) + z \in K_2$ because $w = v - u \in K_2$. This implies $K_1 \subseteq K_2$. The reverse inclusion can be shown similarly.

Definition

Let D be a non-empty affine subspace in a linear space L . The **dimension** of D (denoted by $\dim(D)$) is the dimension of the unique subspace K of L such that $D = t_u(K)$ for some translation t_u of L .

The **dimension of a convex set C** is the dimension of the affine space $K_{\text{aff}}(C)$.

Since \emptyset is an affine subspace of L and there is no subspace of L that can be translated into \emptyset , the dimension of \emptyset is set through the special definition $\dim(\emptyset) = -1$.

Let D, E be two affine subspaces in a linear space L . The sets D, E are **parallel** if $E = t_a(D)$, for some translation t_u of L . In this case we write $D \parallel E$.

It is easy to see that “ \parallel ” is an equivalence relation on the set of affine subspaces of a linear space L . Furthermore, each equivalence class contains exactly one subspace of L .

Affine Subspaces and Linear Systems

with solving linear systems.

Theorem

Let $A \in \mathbb{R}^{m \times n}$ and let $\mathbf{b} \in \mathbb{R}^m$. The set $S = \{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{b}\}$ is an affine subset of \mathbb{R}^n . Conversely, every affine subset of \mathbb{R}^n is the set of solutions of a system of the form $A\mathbf{x} = \mathbf{b}$.

Proof

It is immediate that the set of solutions of a linear system is affine. Conversely, let S be an affine subset of \mathbb{R}^n and let L be the linear subspace such that $S = \mathbf{u} + L$. Let $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ be a basis of L^\perp . We have

$$L = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{a}'_i \mathbf{x} = 0 \text{ for } 1 \leq i \leq m\} = \{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}\},$$

where A is a matrix whose rows are $\mathbf{a}'_1, \dots, \mathbf{a}'_m$. By defining $\mathbf{b} = A\mathbf{u}$ we have

$$S = \{\mathbf{u} + \mathbf{x} \mid A\mathbf{x} = \mathbf{0}\} = \{\mathbf{y} \in \mathbb{R}^n \mid A\mathbf{y} = \mathbf{b}\}.$$

Definition

A subset $U = \{x_1, \dots, x_n\}$ of a real linear space L is **affinely dependent** if $0_L = a_1x_1 + \dots + a_nx_n$, at least one of the numbers a_1, \dots, a_n is nonzero, and $\sum_{i=1}^n a_i = 1$. If no such affine combination exists, then x_1, \dots, x_n are **affinely independent**.

Theorem

Let $U = \{\mathbf{x}_1, \dots, \mathbf{x}_n\}$ be a finite subset of a real linear space L . The set U is affinely independent if and only if the set $V = \{\mathbf{x}_1 - \mathbf{x}_n, \dots, \mathbf{x}_{n-1} - \mathbf{x}_n\}$ is linearly independent.

Proof

Suppose that U is affinely independent but V is linearly dependent; that is, $0_L = b_1(x_1 - x_n) + \cdots + b_{n-1}(x_{n-1} - x_n)$ such that not all numbers b_i are 0. This implies $b_1x_1 + \cdots + b_{n-1}x_{n-1} - \left(\sum_{i=1}^{n-1} b_i\right)x_n = \mathbf{0}$, which contradicts the affine independence of U .

Proof (cont'd)

Conversely, suppose that V is linearly independent but U is not affinely independent. In this case, $0_L = a_1x_1 + \cdots + a_nx_n$ such that at least one of the numbers a_1, \dots, a_n is nonzero and $\sum_{i=1}^n a_i = 0$. This implies $a_n = -\sum_{i=1}^{n-1} a_i$, so $0_L = a_1(x_1 - x_n) + \cdots + a_{n-1}(x_{n-1} - x_n)$. Observe that at least one of the numbers a_1, \dots, a_{n-1} must be distinct from 0 because otherwise we would have $a_1 = \cdots = a_{n-1} = a_n = 0$. This contradicts the linear independence of V , so U is affinely independent.

Definition

The subset $U = \{\mathbf{x}_1, \dots, \mathbf{x}_n\}$ is in **general position** if its points are affinely independent, or equivalently, if the set $V = \{\mathbf{x}_1 - \mathbf{x}_n, \dots, \mathbf{x}_{n-1} - \mathbf{x}_n\}$ is linearly independent.

Corollary

The maximal size of an affinely independent set of vectors in \mathbb{R}^n is $n + 1$.

Proof: Since the maximal size of a linearly independent set in \mathbb{R}^n is n , it follows that the maximal size of an affinely independent set in \mathbb{R}^n is $n + 1$.

Example

Let $\mathbf{x}_1, \mathbf{x}_2$ be vectors in \mathbb{R}^2 . The line that passes through \mathbf{x}_1 and \mathbf{x}_2 consists of all vectors \mathbf{x} such that $\mathbf{x} - \mathbf{x}_1$ and $\mathbf{x} - \mathbf{x}_2$ are collinear; that is, $a(\mathbf{x} - \mathbf{x}_1) + b(\mathbf{x} - \mathbf{x}_2) = \mathbf{0}$ for some $a, b \in \mathbb{R}$ such that $a + b \neq 0$. Thus, we have $\mathbf{x} = a_1\mathbf{x}_1 + a_2\mathbf{x}_2$, where $a_1 = \frac{a}{a+b}$, $a_2 = \frac{b}{a+b}$ and $a_1 + a_2 = 1$, so \mathbf{x} is an affine combination of \mathbf{x}_1 and \mathbf{x}_2 . On other hand, the segment of line contained between \mathbf{x}_1 and \mathbf{x}_2 is consists of convex combinations of \mathbf{x}_1 and \mathbf{x}_2 .

Theorem

The intersection of any collection of convex sets in a real linear space is a convex set.

The intersection of any collection of affine subspaces in a real linear space is an affine subspace.

Proof

Let $\mathcal{C} = \{C_i \mid i \in I\}$ be a collection of convex sets and let $C = \bigcap \mathcal{C}$. Suppose that $x_1, \dots, x_k \in C$, $a_i \geq 0$ for $1 \leq i \leq k$, and $a_1 + \dots + a_k = 1$. Since $x_1, \dots, x_k \in C_i$, it follows that $a_1 x_1 + \dots + a_k x_k \in C_i$ for every $i \in I$. Thus, $a_1 x_1 + \dots + a_k x_k \in C$, which proves the convexity of C . The argument for the affine subspaces is similar.

Corollary

The families of convex sets and non-empty affine subspaces in a real linear space L are closure systems.

Proof: This statement follows immediately by observing that L itself is both a convex set and an affine subspace.

The **convex hull** (or the **convex closure**) of a subset U of L is the intersection $\mathbf{K}_{\text{conv}}(U)$ of all closed sets that contain the set U . Similarly, the **affine hull** of U , denoted by $\mathbf{K}_{\text{aff}}(U)$, is the intersection of all affine sets that contain U .

It is immediate that $\mathbf{K}_{\text{conv}}(U)$ consists of all convex combinations of elements of U , $\mathbf{K}_{\text{aff}}(U)$ consists of all affine combinations of the same elements, and

$$\mathbf{K}_{\text{conv}}(U) \subseteq \mathbf{K}_{\text{aff}}(U) \subseteq \langle U \rangle$$

because each convex combination is also an affine combination and each affine combination is a linear combination.

The dimension $\dim(D)$ of an affine subspace D is the dimension of the unique subspace K of L such that $D = t_u(K)$ for some translation t_u .

Theorem

For an affine subspace D of \mathbb{R}^n we have $\dim(D) = m - 1$ if and only if m is the largest non-negative integer such that there exists an affinely independent set of m elements of D .

Proof

Suppose that $\dim(D) = m - 1$ and $D = t_{\mathbf{u}}(K)$, where K is a subspace of \mathbb{R}^n of dimension $m - 1$ and $\mathbf{u} \notin K$. Let $Y = \{\mathbf{y}_1, \dots, \mathbf{y}_{m-1}\}$ be a basis of K .

The set that consists of m vectors of D

$$\mathbf{x}_1 = \mathbf{u} + \mathbf{y}_1, \dots, \mathbf{x}_{m-1} = \mathbf{u} + \mathbf{y}_{m-1}, \mathbf{x}_m = \mathbf{u}$$

is affinely independent because the set $\{\mathbf{x}_1 - \mathbf{x}_m, \dots, \mathbf{x}_{m-1} - \mathbf{x}_m\}$ is linearly independent.

There is no affinely independent set in D that consists of more than m point because this would entail the existence in K of a basis that consists of more than $m - 1$ vectors.

Let S be a non-empty subset of \mathbb{R}^n . If $\mathbf{0}_n \in \mathbf{K}_{\text{aff}}(S)$, it follows that $\mathbf{K}_{\text{aff}}(S)$ is a subspace of \mathbb{R}^n that coincides with the subspace $\langle S \rangle$ generated by S , and $\dim(\mathbf{K}_{\text{aff}}(S)) = \dim(\langle S \rangle)$.

Theorem

(Stone's Theorem) *Let L be a real linear space and let A, B be two disjoint convex subsets of L . There exists a partition $\pi = \{C, D\}$ of L such that C and D are convex, $A \subseteq C$ and $B \subseteq D$.*

Proof

Let $\mathcal{E} = \{E \in \mathcal{P}(L) \mid E \text{ is convex}, A \subseteq E, B \cap E = \emptyset\}$. Clearly, $\mathcal{E} \neq \emptyset$ because $A \in \mathcal{E}$. The collection \mathcal{E} is partially ordered by set inclusion, so by Zorn's Lemma, it contains a maximal element C , which is clearly convex and disjoint from B . We need to show only that $D = L - C$ is convex. If $D = B$, D is convex and the argument is complete. Therefore, we assume that $B \subset D$, so the set $D - B$ is non-empty.

Proof (cont'd)

If D were not convex, then we would have x, z in D such that $[x, z] \cap C \neq \emptyset$, so we would have $y \in (x, z) \cap C$, that is $y = (1 - c)x + cz$ for some $c \in (0, 1)$.

Note that we cannot have both $x \in B$ and $z \in B$ because this would imply that $C \cap B \neq \emptyset$. Thus, **at least one of x and z must not belong to B .**

Suppose for now that neither x nor z belong to B .

Proof (cont'd)

We claim that there is a point $p \in C$ such that $(p, x) \cap B \neq \emptyset$ and a point $q \in C$ such that $(q, z) \cap B \neq \emptyset$. Equivalently, if for all $p \in C$, $(p, x) \cap B = \emptyset$, or for all $q \in C$, $(q, z) \cap B = \emptyset$, then C is not a maximal convex set that contains A and is disjoint from B . Assume that for all $p \in C$, $(p, x) \cap B = \emptyset$. Then, $C \subseteq \mathbf{K}_{\text{conv}}(\{x\} \cup C)$. Since $x \notin B$, $\mathbf{K}_{\text{conv}}(\{x\} \cup C)$ is disjoint from B , which contradicts the maximality of C . Therefore, there exists $p \in C$ such that $(p, x) \cap B \neq \emptyset$. Similarly, there exists $q \in C$ such that $(q, z) \cap B \neq \emptyset$.

Proof (cont'd)

Let $u \in (p, x) \cap B$ and let $v \in (q, z) \cap B$. We have

$$u = (1 - a)p + ax \text{ and } v = (1 - b)q + bz$$

for some $a, b \in (0, 1)$. Since

$$\begin{aligned} x &= \frac{1}{a}u - \frac{1-a}{a}p, \\ z &= \frac{1}{b}v - \frac{1-b}{b}q, \end{aligned}$$

we have

$$y = \frac{1-c}{a}u - \frac{(1-c)(1-a)}{a}p + \frac{c}{b}v - \frac{c(1-b)}{b}q,$$

or, equivalently

$$\frac{1-c}{a}u + \frac{c}{b}v = y + \frac{(1-c)(1-a)}{a}p + \frac{c(1-b)}{b}q. \quad (1)$$

Proof (cont'd)

Observe that

$$\frac{1-c}{a} + \frac{c}{b} = 1 + \frac{(1-c)(1-a)}{a} + \frac{c(1-b)}{b}. \quad (2)$$

Let k be the value of either side of the equality. Since the coefficients that occur in both sides of Equality (1) are non-negative, by dividing both sides of this equality by k we obtain a convex combination of u and v equal to a convex combination of y, p and q . This contradicts that C and B are disjoint. Therefore, D is convex.

Suppose now that $x \in B$ and $z \notin B$. The role played previously by u will be played by x and the previous argument is applicable with $x = u$.

Theorem

Every affine subset S of \mathbb{R}^n is the intersection of a finite collection of hyperplanes.

Proof: S can be written as $S = \{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{b}\}$, where $A \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^m$. Therefore, $\mathbf{x} \in S$ if and only if $\mathbf{a}_i' \mathbf{x} = b_i$, where \mathbf{a}_i is the i^{th} row of A . Thus, $S = \bigcap_{i=1}^m H_{\mathbf{a}_i, b_i}$.

The class of convex set is closed with respect to scalar multiplications and translations. In other words, it is immediate that if C is a subset of a real linear space, then $h_r(C) = rC = \{rx \mid x \in C\}$ is a convex set for $r \in \mathbb{R}$; also, for $b \in L$ the set $t_b(C) = C + b = \{x + b \mid x \in C\}$ is convex. The **Minkowski sum** of two subsets C_1, C_2 of \mathbb{R}^n is the set

$$C_1 + C_2 = \{\mathbf{x}_1 + \mathbf{x}_2 \mid \mathbf{x}_1 \in C_1, \mathbf{x}_2 \in C_2\}.$$

Theorem

If C_1, C_2 are convex subsets of a real linear space L , their Minkowski sum $C_1 + C_2$ is a convex subset of L .

Proof

Let $x, y \in C_1 + C_2$. We have $x = x_1 + x_2$ and $y = y_1 + y_2$, where $x_1, y_1 \in C_1$ and $x_2, y_2 \in C_2$. Therefore, for $c \in [0, 1]$ we have

$$\begin{aligned}(1-a)x + ay &= (1-a)(x_1 + x_2) + a(y_1 + y_2) \\ &= (1-a)x_1 + ay_1 + (1-a)x_2 + ay_2 \in C_1 + C_2,\end{aligned}$$

because $(1-a)x_1 + ay_1 \in C_1$ and $(1-a)x_2 + ay_2 \in C_2$ because of the convexity of C_1 and C_2 .

If C_1, \dots, C_m are convex sets and r_1, \dots, r_m then the set $r_1 C_1 + \dots + r_m C_m$ is convex.

Theorem

Let C be a convex subset of a real linear space L . If $r_1, r_2 \in \mathbb{R}_{\geq 0}$, then we have

$$(r_1 + r_2)C = r_1 C + r_2 C.$$

Proof

If at least one of r_1, r_2 is 0 the equality obviously holds; therefore, assume that both r_1 and r_2 are positive.

Let $z \in r_1 C + r_2 C$. There exists $x, y \in C$ such that $z = r_1 x + r_2 y$, and therefore,

$$z = (r_1 + r_2) \left(\frac{r_1}{r_1 + r_2} x + \frac{r_2}{r_1 + r_2} y \right).$$

Since C is convex, $\frac{r_1}{r_1 + r_2} x + \frac{r_2}{r_1 + r_2} y \in C$, which implies $z \in (r_1 + r_2)C$, so $r_1 C + r_2 C \subseteq (r_1 + r_2)C$. The reverse inclusion is immediate and makes no use of the convexity of C .

Definition

Let L be a real linear space. A **cone** in L is a non-empty set $C \subseteq L$ such that $x \in C$ and $a \in \mathbb{R}_{\geq 0}$ imply $ax \in C$.

Example

Let L be a real linear space and let S be a non-empty subset of L . The set

$$C_S = \{ax \mid a \geq 0 \text{ and } x \in S\}$$

is cone contained by every other cone that contains S .

Example

The set $(\mathbb{R}_{\geq 0})^n$ is a pointed cone.

Theorem

Let L be a real linear space and let $C \in L$ be a cone. C is convex if and only if $x + y \in C$ for $x, y \in \mathbb{R}$.

Proof

Let C be a convex cone. If $x, y \in C$ and $a \in (0, 1)$, then $\frac{1}{a}x \in C$ and $\frac{1}{1-a}y \in C$. Therefore, by convexity we have

$$x + y = a\frac{1}{a}x + (1-a)\frac{1}{1-a}y \in C.$$

Conversely, let C be a cone such that $x, y \in C$ imply $x + y \in C$. For $u, v \in C$ and $a \in [0, 1]$ let $z_a = au + (1-a)v$. Since C is a cone, $au \in C$ and $(1-a)v \in C$, hence $z_a = au + (1-a)v \in C$. Therefore, C is convex.

Example

Let U be a non-empty subset of a real linear space L . The set of all non-negative combinations of U is a convex cone that is included in every convex cone that contains U .

Theorem

The intersection of any collection of cones (convex cones) in a real linear space L is a cone (a convex cone).

Corollary

The families of cones (convex cones) in a real linear space L is a closure system.

Proof: This statement follows immediately by observing that \mathbb{R}^n itself is cone (a convex cone).

We denote the closure operator corresponding to the family of cones by K_{cone} .

Theorem

Let S be a non-empty subset of a real linear space L . We have
 $\mathbf{K}_{\text{cone}}(S) = \{ax \mid a \geq 0 \text{ and } x \in S\}.$

Proof: Since $\{ax \mid a \geq 0 \text{ and } x \in S\}$ is a cone that contains S ,
 $\mathbf{K}_{\text{cone}}(S) \subseteq \{ax \mid a \geq 0 \text{ and } x \in S\}.$

Conversely, since $S \subseteq \mathbf{K}_{\text{cone}}(S)$, if $x \in S$ it follows that $ax \in \mathbf{K}_{\text{cone}}(S)$ for every $a \geq 0$, so $\{ax \mid a \geq 0 \text{ and } x \in S\} \subseteq \mathbf{K}_{\text{cone}}(S).$

Definition

Let C be a non-empty convex subset of a real linear space L . An **extreme point** of C is a point $x \in C$ such that if $x \in [u, v]$ and $u, v \in C$, then $u = v = x$.

Theorem

Let C be a non-empty convex subset of a real linear space L . A point $x \in C$ is an extreme point of C if the set $C - \{x\}$ is convex.

Proof

Suppose that $C - \{x\}$ is a convex set for $x \in C$ and that $x \in [u, v]$, where $u, v \in C$.

If x is distinct from both u and v , then u, v belong to the convex set $C - \{x\}$, which yields the contradiction $x \in C - \{x\}$. Thus, x is an extreme point of C .

Conversely, suppose that x is an extreme point of C . Let $u, v \in C - \{x\}$, so $u \neq x$ and $v \neq x$. If $x \in [u, v]$, we obtain a contradiction since this implies $u = v = x$. Therefore $[u, v] \subseteq C - \{x\}$, so $C - \{x\}$ is convex. The set of extreme points of a convex set C is denoted by $\text{extr}(C)$.

Example

Let $B[\mathbf{x}_0, r]$ be a closed sphere of radius r in \mathbb{R}^n . Each point \mathbf{x} located on the circumference of this sphere, that is, each point \mathbf{x} such that

$\|\mathbf{x}_0 - \mathbf{x}\| = r$ is an extreme point of $B[\mathbf{x}_0, r]$.

Indeed, suppose that $a\mathbf{u} + (1 - a)\mathbf{v} = \mathbf{x}$ for some $a \in (0, 1)$ and

$\|\mathbf{x}_0 - \mathbf{u}\| = \|\mathbf{x}_0 - \mathbf{v}\| = r$. Then, by Supplement ??, we have $\mathbf{u} = \mathbf{v} = \mathbf{x}$.

Example

An open sphere $B(\mathbf{x}_0, r)$ in \mathbb{R}^n has no extreme points for if $\mathbf{x} \in B(\mathbf{x}_0, r)$. Indeed, let \mathbf{u} be a vector such that $\mathbf{u} \neq \mathbf{0}_n$ and let $\mathbf{x}_1 = \mathbf{x} + a\mathbf{u}$ and $\mathbf{x}_2 = \mathbf{x} - a\mathbf{u}$, where $a > 0$. Observe that if $a < \frac{r - \|\mathbf{x} - \mathbf{x}_0\|}{\|\mathbf{u}\|}$, we have

$$\|\mathbf{x}_1 - \mathbf{x}_0\| = \|\mathbf{x} + a\mathbf{u} - \mathbf{x}_0\| \leq \|\mathbf{x} - \mathbf{x}_0\| + a\|\mathbf{u}\| < r$$

and

$$\|\mathbf{x}_2 - \mathbf{x}_0\| = \|\mathbf{x} - a\mathbf{u} - \mathbf{x}_0\| \leq \|\mathbf{x} - \mathbf{x}_0\| + a\|\mathbf{u}\| < r,$$

and we have both $\mathbf{x}_1 \in B(\mathbf{x}_0, r)$ and $\mathbf{x}_2 \in B(\mathbf{x}_0, r)$. Since $\mathbf{x} = \frac{1}{2}\mathbf{x}_1 + \frac{1}{2}\mathbf{x}_2$, \mathbf{x} is not an extreme point.

Example

The extreme points of the cube $[0, 1]^n$ are all its 2^n “corners”
 $(a_1, \dots, a_n) \in \{0, 1\}^n$.

Definition

Let C be a convex set in a real linear space L . A convex subset F of C is a **face** of C if for every open segment $(u, v) \subseteq C$ such that at least one of u, v is not in F we have $(u, v) \cap F = \emptyset$.

If $F \neq C$, we say that F is a **proper face** of C .

A **k -face** of C is a face F of C such that $\dim(F) = k$.

A convex subset F is a face of C if $u, v \in C$ and $(u, v) \cap F \neq \emptyset$ implies $u \in F$ and $v \in F$, which is equivalent to $[u, v] \subseteq F$.

Note that if $F = \{x\}$ is a face of C if and only if $x \in \text{extr}(C)$. A convex subset C is a face of itself.

Theorem

If F is a face of a convex set C , then $F = \mathbf{K}_{\text{aff}}(F) \cap C$.

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If F is a face of a convex set C , then $F = \mathbf{K}_{\text{aff}}(F) \cap C$.

Proof

If $z \in \mathbf{K}_{\text{aff}}(F) \cap C$, we have $x = a_1 y_1 + \cdots + a_k y_k$, where $\sum_{i=1}^k a_i = 1$ and $y_1, \dots, y_k \in F$. If all a_i are non-negative, then it is immediate that $x \in F$. Otherwise, let $b = -\sum \{a_i \mid a_i < 0\}$ and let

$$\begin{aligned} u &= \frac{1}{1+b} \sum \{a_i y_i \mid a_i \geq 0\} \\ v &= -\frac{1}{b} \sum \{a_i y_i \mid a_i < 0\}. \end{aligned}$$

We have $x \in C$, $v \in C$, and

$$u = \frac{1}{1+b} x + \frac{b}{1+b} v \in [x, v] \cap F.$$

Since F is a face, we have $u \in F$. Thus, $\mathbf{K}_{\text{aff}}(F) \cap C \subseteq F$. The reverse inclusion is immediate.

