

# Convex

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## 1 Affine sets

$$\mathbf{x} = (\xi_1, \dots, \xi_n) \in \mathbb{R}^n$$

$$\langle x, x^* \rangle = \sum_{i=1}^n \xi_i \cdot \xi_i^*$$

$A : m \times n$  real matrix and linear operator from  $\mathbb{R}^n$  to  $\mathbb{R}^m$

$$f : \mathbb{R}^n \rightarrow \mathbb{R}^m : x \mapsto Ax$$

$A^* = \overline{A}^\top$  : adjoint matrix (随伴行列、エルミート転置行列)

(the transpose matrix and the corresponding adjoint linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ )

$$\langle Ax, y^* \rangle = \langle x, A^* y^* \rangle$$

**Definition 1.1** (affine set). A subset  $M$  of  $\mathbb{R}^n$  is called *affine set* if

$$\theta x + (1 - \theta)y \in M, \quad \forall x, y \in M, \forall \theta \in \mathbb{R}$$

The empty set  $\emptyset$  and the space  $\mathbb{R}^n$  itself are affine sets. Also covered by the definition is the case where  $M$  consists of a solitary point. (一点集合も Affine set) An affine set has to contain, along with any two different points, the entire line through those points.

**Theorem 1.1.** *The subspaces of  $\mathbb{R}^n$  are the affine sets which contain the origin.*

**Definition 1.2.** For  $M \subset \mathbb{R}^n$  and  $a \in \mathbb{R}^n$ , the translate(平行移動) of  $M$  by  $a$  is defined to be the set

$$M + a = \{x + a \mid x \in M\}$$

An affine set  $M$  is said to be parallel to an affine set  $L$  if  $L = M + a$  for some  $a$ .

**Theorem 1.2.** *Each non-empty affine set  $M$  is parallel to a unique subspace  $L$ . This  $L$  is given by*

$$L = C - C = \{x - y \mid x, y \in C\}$$

The dimension of a non-empty affine set is defined as the dimension of the subspace parallel to it. (The dimension of  $\emptyset$  is  $-1$  by convention.)

空でないアフィン集合の次元は、それに平行な部分空間の次元と定義される。 $(\emptyset$  の次元は慣習的に  $-1$ )  
 Affine sets of dimension 0, 1 and 2 are called points, lines and planes, respectively.

**Definition 1.3** (Affine combination). Let  $x_1, \dots, x_n \in \mathbb{R}^n$ , and  $\theta_1, \dots, \theta_n \in \mathbb{R} (n \in \mathbb{N})$ .

$\sum_{i=1}^n \theta_i x_i$  is an affine combination of  $x_1, \dots, x_n$  if  $\sum_{i=1}^n \theta_i = 1$

**Theorem 1.3** (Hyperplane). Given  $\beta \in \mathbb{R}$  and a non-zero  $\mathbf{b} \in \mathbb{R}^n$ , the set

$$H = \{x \in \mathbb{R}^n \mid \langle \mathbf{x}, \mathbf{b} \rangle = \beta\}$$

is a hyperplane in  $\mathbb{R}^n$ .

The vector  $\mathbf{b}$  is called a normal to the hyperplane  $H$ .

Every hyper plane has "two sides," like one's picture of a line in  $\mathbb{R}^2$  or a plane in  $\mathbb{R}^3$ .

**Theorem 1.4.** Given  $\mathbf{b} \in \mathbb{R}^m$  and an  $m \times n$  real matrix  $\mathbf{B}$ , the set

$$M = \{x \in \mathbb{R}^n \mid \mathbf{B}\mathbf{x} = \mathbf{b}\}$$

is an affine set in  $\mathbb{R}^n$ . Moreover, every affine set may be represented in this way.

Let  $\mathbf{b}_i$  is the  $i$ th row of  $\mathbf{B}$ ,  $\beta_i$  is the  $i$ -th component of  $\mathbf{b}$ , and

$$H_i = \{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b}_i \rangle = \beta_i, \}.$$

Then,

$$M = \{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b}_i \rangle = \beta_i, i = 1, \dots, m\} = \bigcap_{i=1}^m H_i.$$

Each  $H_i$  is a hyperplane ( $\mathbf{b}_i \neq \mathbf{0}$ ).

The affine set  $M$  in Theorem can be expressed in terms of the vectors  $b'_1, \dots, b'_n$  which form the columns of  $\mathbf{B}$  by

$$M = \{\mathbf{x} = (\xi_1, \dots, \xi_n) \mid \sum_{i=1}^n \xi_i b'_i = \mathbf{b}\}$$

任意の集合  $S \subset \mathbb{R}^n$  に対し、 $S$  を含む（唯一で）最小のアフィン集合が存在する。これをアフィン包 (affine hull) という。

**Definition 1.4** (affine hull). A subset  $C$  of  $\mathbb{R}^n$  is called *affine hull* if

$$\text{aff } S = \left\{ \sum_{i=1}^n \theta_i x_i \mid x_i \in S (i = 1, \dots, n), \sum_{i=1}^n \theta_i = 1 \right\}.$$

**Definition 1.5.** A set of  $k+1$  points  $b_0, b_1, \dots, b_k$  is said *affinely independent* if the set

$$\begin{aligned}\text{aff}\{b_0, b_1, \dots, b_k\} &= \text{aff}\{b_0 - b_0, b_1 - b_0, \dots, b_k - b_0\} + b_0 \\ &= \text{aff}\{0, b_1 - b_0, \dots, b_k - b_0\} + b_0 \\ &= L + b_0 \quad \because L = \text{aff}\{0, b_1 - b_0, \dots, b_k - b_0\}\end{aligned}$$

is  $k$ -dimensional.

定理 1.1 より、 $L$  は  $b_1 - b_0, \dots, b_k - b_0$  を含む最小の部分集合と同じになる。 $L$  の次元が  $k$  になる必要十分条件は  $b_1 - b_0, \dots, b_k - b_0$  が、線形独立になることである。従って、 $b_0, b_1, \dots, b_k$  がアフィン独立である必要十分条件は  $b_1 - b_0, \dots, b_k - b_0$  が線形独立であることである。

よって  $x_0, x_1, \dots, x_k$  がアフィン独立であれば、 $x \in \text{aff}\{b_0, b_1, \dots, b_k\}$  は次のように表せる。

$$x = \sum_{i=0}^k \theta_i x_i, \quad \sum_{i=0}^k \theta_i = 1.$$

**Definition 1.6** (affine transform).  $T: x \rightarrow Tx$  from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  is called *affine transformation* if

$$T((1-\lambda)x + \lambda y) = (1-\lambda)Tx + \lambda Ty, \quad \forall x, y \in \mathbb{R}^n, \forall \lambda \in \mathbb{R}$$

## 2 Convex Sets and Cones

**Definition 2.1** (convex set). A subset  $C$  of  $\mathbb{R}^n$  is called *convex* if

$$\theta x + (1-\theta)y \in C, \quad \forall x, y \in C, \forall \theta \in [0, 1]$$

All affine sets are convex.

**Definition 2.2** (convex combination). An affine combination with a coefficient  $\theta_i \in [0, 1]$  is called a *convex combination*, i.e.

let  $x_1, \dots, x_n \in \mathbb{R}^n$ , and  $\theta_1, \dots, \theta_n \in \mathbb{R} (n \in \mathbb{N})$ .  $\sum_{i=1}^n \theta_i \cdot x_i$  is an *convex combination* of  $x_1, \dots, x_n$  if

$$\sum_{i=1}^n \theta_i x_i \quad (\theta_i \in [0, 1] (i = 1, \dots, n)), \quad \sum_{i=1}^n \theta_i = 1,$$

**Definition 2.3** (Half-spaces(半空間)). For any non-zero  $\mathbf{b} \in \mathbb{R}^n$  and any  $\beta$  in  $\mathbb{R}$ , the sets

$$\{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b} \rangle \leq \beta\}, \quad \{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b} \rangle \geq \beta\}$$

are called *closed half-spaces*. The sets

$$\{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b} \rangle < \beta\}, \quad \{\mathbf{x} \mid \langle \mathbf{x}, \mathbf{b} \rangle > \beta\}$$

are called *open half-spaces*. All four sets are plainly non-empty and convex.

**Theorem 2.1.** *The intersection of an arbitrary collection of convex sets is convex.*

A set which can be expressed as the intersection of finitely many closed half spaces of  $\mathbb{R}^n$  is called a *polyhedral convex set*. (凸多面体)

$H = \{x \in \mathbb{R}^n \mid \langle x, b \rangle = \beta\}$ ,  $a \in H$  とするとき、半空間  $(\{x \in \mathbb{R}^n \mid \langle x, b \rangle \leq \beta\})$  は

$$C = \{x \in \mathbb{R}^n \mid \langle x - a, b \rangle \leq 0\}$$

と表せる。 $b$  は点  $a$  における法線ベクトルである。

これは点  $a$  から半空間  $(\{x \in \mathbb{R}^n \mid \langle x, b \rangle \leq \beta\})$  上の点  $a$  に向かうベクトルが超平面の法線ベクトル  $b$  と鈍角をなすことを意味している。

よって、半空間  $(\{x \in \mathbb{R}^n \mid \langle x, b \rangle \leq \beta\})$  は  $b$  の反対側に位置する。

一方、半空間  $(\{x \in \mathbb{R}^n \mid \langle x, b \rangle \geq \beta\})$  は  $b$  の同じ側に位置する。

**Corollary 2.1.1.** *Let  $b_i \in \mathbb{R}^n$  and  $\beta_i \in \mathbb{R}$  for  $i \in I$ , where  $I$  is an arbitrary index set, and consider the set*

$$C = \{x \in \mathbb{R}^n \mid \langle x, b_i \rangle \leq \beta_i, \forall i \in I\}.$$

*It is convex.*

**Theorem 2.2.** *A set  $C \subset \mathbb{R}^n$  is convex if and only if it contains all the convex combinations of its elements.*

( $C \subset \mathbb{R}^n$  が凸集合  $\iff C$  の元からなる全ての凸結合を  $C$  自身が含む)

**Theorem 2.3** (convex hull). *The convex hull of a given set  $X \subset \mathbb{R}^n$  may be defined as the set satisfying any one (and hence all) of the following equivalence conditions.*

1. *The (unique) minimal convex set containing  $X$ . ( $X$  を含む (唯一の) 最小の凸集合)*
2. *The intersection of all convex sets containing  $X$ .*

$$\text{conv}X = \bigcap \{C : C \text{ is convex set, } X \subset C\}$$

3. *The set of all the convex combinations of points in  $X$ . ( $X$  に属する点から得られる凸結合全体の成す集合)*

$$\text{conv}X = \left\{ \sum_{i=1}^n \theta_i x_i \in \mathbb{R}^n \mid \exists m \in \mathbb{N}, \exists x_1, \dots, x_m \in X, \exists \theta_1, \dots, \theta_m \in [0, 1], \sum_{i=1}^m \theta_i = 1 \right\}$$

4. *(Carathéodory's Theorem) The union of all simplices with vertices in  $X$ . ( $X$  に属する点を頂点とする単体全ての合併)*

$$\text{conv}X = \left\{ \sum_{i=1}^{n+1} \theta_i x_i \mid n \in \mathbb{N}, x_i \in X, \theta_i > 0, \sum_{i=1}^{n+1} \theta_i = 1 \right\}$$

A set which is a convex hull of a finite number of points is called a polytope(超多面体).

**Definition 2.4.** A subset  $K$  of  $\mathbb{R}^n$  is called *cone* if it is closed under positive scalar multiplication, i.e.  $\lambda x \in K$  for  $x \in K$  and  $\lambda > 0$ .

**Definition 2.5.** A subset  $K$  of  $\mathbb{R}^n$  is called *convex cone* if it is closed under positive scalar multiplication, i.e.  $\lambda x + (1 - \lambda)y \in K$  for  $x, y \in K$  and  $\lambda > 0$ .

If  $K$  contains the origin, we call it a convex cone.

A convex cone should not necessarily be considered “pointed”. A subspace of  $\mathbb{R}^n$  is a particularly convex cones. The same is true is the open and closed half-space corresponding to a hyperplane through the origin.

Non-negative *orthant* of  $\mathbb{R}^n$  is convex cone.

$$\begin{aligned}\mathbb{R}_+^n &= \{x = (\xi_1, \dots, \xi_n) \in \mathbb{R}^n \mid \xi_i \geq 0 \text{ for } i = 1, \dots, n\} = \{x \in \mathbb{R}^n \mid x \geq 0\} \\ \mathbb{R}_{>0}^n &= \{x = (\xi_1, \dots, \xi_n) \in \mathbb{R}^n \mid \xi_i > 0 \text{ for } i = 1, \dots, n\} = \{x \in \mathbb{R}^n \mid x > 0\}\end{aligned}$$

**Theorem 2.4.** The intersection of an arbitrary collection of convex cones is a convex cone.

**Corollary 2.4.1.** Let  $b_i \in \mathbb{R}^n$  for  $i \in I$ , where  $I$  is an arbitrary index set. Then

$$K = \{x \in \mathbb{R}^n \mid \langle x, b_i \rangle \leq 0, i \in I\}$$

is a convex cone.

### 3 Convex Functions

This note is excluded the value  $f(x) = -\infty$ .

**Definition 3.1.** A function  $f : C \rightarrow \mathbb{R}$ , where  $C$  is a convex set. Then  $f$  is *convex* on  $C$  if and only if

$$f((1 - \lambda)x + \lambda y) \leq (1 - \lambda)f(x) + \lambda f(y), \quad \forall x, y \in C, \quad \lambda \in [0, 1].$$

Let  $f : \text{convex function}$ ,  $f \in C \neq \emptyset$ . Extending  $f$  to  $\mathbb{R}^n$  by setting  $f(x) = +\infty$  for  $x \notin C \subset \mathbb{R}^n$ . Then the above definition is equivalent to the following definition.

**Definition 3.2.** A function  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ , not identically  $+\infty$ , is convex if and only if

$$f((1 - \lambda)x + \lambda y) \leq (1 - \lambda)f(x) + \lambda f(y), \quad \forall x, y \in \mathbb{R}^n, \quad \lambda \in [0, 1].$$

$f : \mathbb{R}^n \rightarrow \mathbb{R} : \text{concave} \Leftrightarrow -f$  is convex.

**Definition 3.3.**  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ ,  $C \subset \mathbb{R}^n$ . We define epigraph of  $f$  as

$$\text{epi} f := \{(x, y) \mid x \in C, y \in \mathbb{R}, y \geq f(x)\}.$$

Note that  $f(x) = +\infty$  for  $x \notin C$ .

In other words,

$f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ , not identically equal to  $+\infty$ . We define epigraph of  $f$  is the non-empty set

$$\text{epi} f := \{(\mathbf{x}, y) \mid \mathbf{x} \in \mathbb{R}^n, y \in \mathbb{R}, y \geq f(\mathbf{x})\}.$$

**Definition 3.4.** The *effective domain* of function  $f$  on  $C$  is the set

$$\text{dom} f = \{\mathbf{x} \in \mathbb{R}^n \mid \exists y, (\mathbf{x}, y) \in \text{epi} f\} = \{\mathbf{x} \in \mathbb{R}^n \mid f(\mathbf{x}) < +\infty\}.$$

**Definition 3.5.** A convex function  $f$  is said to be *proper* if  $f(x) < +\infty$  at least one  $x \in C$  and  $f(x) > -\infty$  for all  $x \in C$ .

In other words,  $f$  is *proper* if its epigraph is non-empty and does not contain “vertical lines”.

Let  $f : \text{convex function}$ ,  $C \subset \mathbb{R}^n$  is non-empty convex set.

$f \in C$  is proper  $\Leftrightarrow$

$\text{dom} f = C$  where  $f \in C$  is finite.

$f \in \mathbb{R}^n$  is proper  $\Leftrightarrow$

$$f \in \mathbb{R}^n \text{ is convex function and } f \in C \text{ is finite.} \Rightarrow f(x) = \begin{cases} f(x) & (x \in C) \\ +\infty & (x \notin C) \end{cases}$$

In this note, we excluded the value  $f(x) = -\infty$ , so from the definition 3.2, every convex function  $f$  is “proper”.

**Theorem 3.1.**  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ .

$f$  is a convex function if and only if  $\text{epi} f$  is convex set.

**Theorem 3.2.** Let  $f : \mathbb{R}^n \rightarrow (-\infty, +\infty]$ ,  $\lambda_i \geq 0$  and  $\sum_{i=0}^k \lambda_i = 1$ . Then  $f$  is convex if and only if

$$f\left(\sum_{i=0}^k \lambda_i x_i\right) \leq \sum_{i=0}^k \lambda_i f(x_i).$$

Neighborhood of  $x'$  is defined by

$$B(x', \delta) := \{\bar{x} \mid d(x', \bar{x}) \leq \delta\}.$$

$$V \in \mathcal{N}(x') := \text{the collection of all neighborhoods of } x'.$$

**Definition 3.6.** (lower limits) Let  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  and let  $x'$  is a limit point of  $f$ . Then the lower limit of function  $f$  is defined by

$$\begin{aligned} \liminf_{x \rightarrow x'} f(x) &= \lim_{\delta \searrow 0} \left[ \inf_{x \in B(x', \delta)} f(x) \right] \\ &= \sup_{\delta > 0} \left[ \inf_{x \in B(x', \delta)} f(x) \right] = \sup_{V \in \mathcal{N}(x')} \left[ \inf_{x \in V} f(x) \right]. \end{aligned}$$

**Definition 3.7.** (lower semi-continuous) Let  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  and let  $x'$  is a limit point of  $f$ . Then  $f$  is lower semi-continuous at  $x'$  if and only if

$$\liminf_{x \rightarrow x'} f(x) \geq f(x'), \text{ or } \liminf_{x \rightarrow x'} f(x) = f(x')$$

**Definition 3.8.** (level set) The lower level sets  $lev_{\leq \alpha} f$  is defined by

$$lev_{\leq \alpha} f := \{x \in \mathbb{R}^n \mid f(x) \leq \alpha\}$$

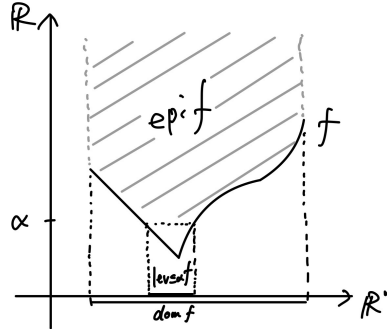


图 1 Level set

**Proposition 3.3.** Let arbitrary function  $f : \mathbb{R}^n \rightarrow [-\infty, +\infty]$ , the following are equivalent:

- (i)  $f$  is lower semi-continuous at all  $x' \in \mathbb{R}^n$ ;
- (ii)  $lev_{\leq \alpha} f := \{x^{s'} \in \mathbb{R}^n \mid f(x') \leq \alpha\}$  is closed for every  $\alpha$ ;
- (iii)  $epi f$  is a closed set in  $\mathbb{R}^{n+1}$ .

**Definition 3.9.** (level boundedness) A function  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  is *lower level bounded* if the set  $lev_{\leq \alpha} f$  is bounded (possibly empty) for every  $\alpha \in \mathbb{R}$ .

**Theorem 3.4.** (conditions  $\inf f = \min f$ ) Suppose  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  is lower semi-continuous, level-bounded and proper. Then the set  $\operatorname{argmin} f$  is nonempty and compact and the value  $\inf f$  is finite. In other words,  $\inf f = \min f$ .

**Corollary 3.4.1.** (lower bounds) Suppose  $f : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  is lsc and proper. Then it is bounded from below on each bounded subset of  $\mathbb{R}^n$ . Actually, it is minimum with respect to any compact subset of  $\mathbb{R}^n$  that satisfies  $\operatorname{dom} f$ .