

Optimization Methods.

Seminar 2. Convex sets.

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- Objective of the Optimization Methods course
- General optimization problem statement
- Examples of optimization problems:
 - linear programming
 - least squares problem
 - convex optimization problems
- Why convex optimization problems are good?

Affine sets

Affine set

Let A be affine set if for any $x_1, x_2 \in A$ and $\theta \in \mathbb{R}$ point $\theta x_1 + (1 - \theta)x_2 \in A$.

Examples: \mathbb{R}^n , hyperplane, single point.

Affine combination of points

Assume $x_1, \dots, x_k \in G$, then point $\theta_1 x_1 + \dots + \theta_k x_k$ is called affine combination of the points x_1, \dots, x_k if $\sum_{i=1}^k \theta_i = 1$.

Affine hull of set

A set $\left\{ \sum_{i=1}^k \theta_i x_i \mid x_i \in G, \sum_{i=1}^k \theta_i = 1 \right\}$ is called affine hull of set G and is denoted by $\mathbf{aff}(G)$.

Claim 1

A set G is affine if and only if it contains all affine combinations of points from G .

Claim 2

A set G is affine if and only if it can be represented in the form $G = \{\mathbf{x} | \mathbf{Ax} = \mathbf{b}\}$.

Convex set

Convex set

Let C be a convex set if

$$\forall x_1, x_2 \in C, \theta \in [0, 1] \rightarrow \theta x_1 + (1 - \theta)x_2 \in C.$$

\emptyset and $\{x_0\}$ are also convex by definition.

Examples: \mathbb{R}^n , affine set, half-open segment, segment.

Convex combination of points

Assume $x_1, \dots, x_k \in G$, then point $\theta_1 x_1 + \dots + \theta_k x_k$ is called convex combination of points x_1, \dots, x_k if $\sum_{i=1}^k \theta_i = 1, \theta_i \geq 0$.

Convex hull

A set $\left\{ \sum_{i=1}^k \theta_i x_i \mid x_i \in G, \sum_{i=1}^k \theta_i = 1, \theta_i \geq 0 \right\}$ is called convex combination of a set G and is denoted by **conv**(G).

Operations that preserve convexity

- Intersection of any number (finite or infinite) of convex sets is convex set.
- Image of convex set under affine function is convex function
- Linear combination of convex sets is convex
- Direct product of convex sets is convex

Examples

Check if the following sets are affine and/or convex:

- Half-space: $\{\mathbf{x} | \mathbf{a}^T \mathbf{x} \leq c\}$
- Polyhedron: $\{\mathbf{x} | \mathbf{A}\mathbf{x} \preceq \mathbf{b}, \mathbf{C}\mathbf{x} = \mathbf{0}\}$
- A ball induced by norm in \mathbb{R}^n :
 $B(r, \mathbf{x}_c) = \{\mathbf{x} \mid \|\mathbf{x} - \mathbf{x}_c\| \leq r\}$
- Ellipsoid: $\mathcal{E}(\mathbf{x}_c, \mathbf{P}, r) = \{\mathbf{x} \mid (\mathbf{x} - \mathbf{x}_c)^T \mathbf{P}^{-1} (\mathbf{x} - \mathbf{x}_c) \leq r\}$
- A set of symmetric and positive-definite matrices:
 $\mathbf{S}_+^n = \{\mathbf{X} \in \mathbb{R}^{n \times n} \mid \mathbf{X}^T = \mathbf{X}, \mathbf{X} \succeq \mathbf{0}\}$
- $\{\mathbf{X} \in \mathbb{R}^{n \times n} \mid \text{Tr}(\mathbf{X}) = \text{const}\}$
- Hyperbolic set: $\{\mathbf{x} \in \mathbb{R}_+^n \mid \prod_{i=1}^n x_i \geq 1\}$

Cone

Cone (convex)

Let a set C be a cone (convex cone), if

$$\forall x \in C, \theta \geq 0 \rightarrow \theta x \in C$$

$$(\forall x_1, x_2 \in C, \theta_1, \theta_2 \geq 0 \rightarrow \theta_1 x_1 + \theta_2 x_2 \in C)$$

Examples: \mathbb{R}^n , affine set with 0, half-open segment.

Conical (non-negative) combination of points

Assume $x_1, \dots, x_k \in G$, then point $\theta_1 x_1 + \dots + \theta_k x_k$ is called conic (non-negative) combination of points x_1, \dots, x_k if $\theta_i \geq 0$.

Conical hull

A set $\left\{ \sum_{i=1}^k \theta_i x_i \mid x_i \in G, \theta_i \geq 0 \right\}$ is called conical hull of a set G and is denoted by **cone**(G).

Examples

- \mathbf{S}_+^n
- Normal cone: $\{(\mathbf{x}, t) \in \mathbb{R}^{n+1} \mid \|\mathbf{x}\| \leq t\}$
In case of ℓ_2 norm it is called second-order cone or Lorentz cone
- Some special cases

Recap

- Affine set
- Convex set
- Cone
- Methods to check properties of given set