Optimization Methods. Seminar 2. Convex sets.

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Reminder

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

Affine hull of set

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

Claims

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 $G = \{\mathbf{x} | \mathbf{A}\mathbf{x} = \mathbf{b}\}$.

Convex set

Convex set

Let C be a convex set if

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

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

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

Convex combination of points

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

Convex hull

A set $\left\{ \sum\limits_{i=1}^{k} \theta_i x_i \mid x_i \in G, \sum\limits_{i=1}^{k} \theta_i = 1, \theta_i \geq 0 \right\}$ 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: $\{\mathbf{x} | \mathbf{a}^\top \mathbf{x} \le c\}$
- Polyhedron: $\{\mathbf{x} | \mathbf{A}\mathbf{x} \preceq \mathbf{b},\ \mathbf{C}\mathbf{x} = 0\}$
- A ball induced by norm in $\mathbb{R}^n$: $B(r, x_c) = \{x \mid \|x - x_c\| \le r\}$
- Ellipsoid: $\mathcal{E}(x_c, \mathbf{P}, r) = \{x \mid (x - x_c)^\top \mathbf{P}^{-1}(x - x_c) \le r\}$
- A set of symmetric and positive-definite matrices: $\mathbf{S}_+^n = \{\mathbf{X} \in \mathbb{R}^{n \times n} \mid \mathbf{X}^\top = \mathbf{X},\ \mathbf{X} \succeq 0\}$
- $\{\mathbf{X} \in \mathbb{R}^{n \times n} \mid \mathrm{Tr}(\mathbf{X}) = const\}$
- Hyberbolic set: $\{\mathbf{x} \in \mathbb{R}_+^n \mid \prod_{i=1}^{n} x_i \ge 1\}$

Cone

Cone (convex)

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

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

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

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

Conical hull

A set $\left\{ \sum\limits_{i=1}^{k} \theta_i x_i \mid x_i \in G, \theta_i \geq 0 \right\}$ 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