

Part I

Introduction to Optimization

1 Introduction to Optimization

Optimization is a branch of mathematics. In practice, one starts from a real-world problem, models it, and solves it mathematically either analytically (as an optimization problem) or numerically (as a mathematical program).

1.0.1 Framework

An optimization problem consists of, given a function $f : S \rightarrow \mathbb{R}$ finding:

- 1) Its minimum value v (or maximum) over the set S .
- 2) A point $x_0 \in S$ at which this minimum (or maximum) is attained, i.e., i.e. $f(x_0) = v$.

1.0.2 Vocabulary

- f is the objective function
 - v is the optimal value
 - x_0 is the optimal solution
 - $S = \{\text{the feasible set (the set of feasible solutions to the problem)}\}$
 - Problem formulation: $\min_{x \in S} f(x)$ resp. $\max_{x \in S} f(x)$

Remark: **Link Between Minimum and Maximum**

The Problem $\max_{x \in S} f(x)$ returns (x_0, v) , then the problem $\min_{x \in S} f(x)$ returns $(x_0, -v)$. Hence the connection. Thus, the search for a maximum can always be reduced to the search for a minimum.

1.0.3 Applications

Optimization is used in many fields:

- In operations research (e.g., transportation problems, economics, inventory management...)
- In numerical analysis (e.g., approximation/solving of linear and nonlinear systems...)
- In control theory (e.g., system modeling, filtering...)

1.1 Different Types of Optimization

1.2 Classification of Optimization Problems

- Linear Optimization

f is a linear function : $f(x) = c, x >$,
 S is defined by affine functions : $ax + b \geq 0$.

- Quadratic Optimization

f is a convex quadratic function: $f(x) = \frac{1}{2} \langle Ax, x \rangle + \langle b, x \rangle$
 A est une matrice symétrique semi-définie positive
 S est défini par des fonctions affines : $ax + b \geq 0$,
-optimisation convexe
 f is a convex function and S is a convex domain
- Differentiable Optimization
 f is a differentiable function
 S is defined by functions (=constraints) differentiable.
- Nondifferentiable Optimization
 example: $f(x) = \max\{f_1(x), \dots, f_m(x)\}$.
-Infinite-dimensional Optimization
 example: Variational problems $J(x) = \int_0^T L(t, x(t), x'(t))dt$ where x belongs to a set of functions $X, x(0) = x_0$ et $x(T) = x_T$.
-Nonlinear Optimization
 We distinguish three types of problems:
-Unconstrained problem: $\min_{x \in \mathbb{R}^n} f(x)$,
-Problem with equality constraints: $\min_{x \in S} f(x)$ with S of the form $S = \{x \in \mathbb{R}^n \mid g_i(x) = 0 \text{ for } i = 1..l\}$ and $g_i: \mathbb{R}^n \rightarrow \mathbb{R}$.
-Problem with inequality constraints: $\min_{x \in S} f(x)$ with S of the form $S = \{x \in \mathbb{R}^n \mid h_i(x) \geq 0 \text{ for } i = 1..l\}$ with $h_i: \mathbb{R}^n \rightarrow \mathbb{R}$.

1.3 Examples of Nonlinear Optimization Problems

A Production Problem

A factory needs n products $a_1, \dots, a_n$ in quantities $x_1, \dots, x_n$ to manufacture a finished product b in quantity $h(x_1, \dots, x_n)$. Let p_i be the unit price of product i and p the selling price of the finished product. We want to calculate the optimal production. The goal is therefore to maximize the function $ph(x_1, \dots, x_n) - \sum_{i=1}^n x_i p_i$.

1.4 Existence of a solution

Weierstrass Theorem: Let f be continuous function on $S \subset \mathbb{R}^n$ with S closed and bounded set. Then $\exists x_0 \in S$, such that $f(x_0) = \min_{x \in S} f(x)$, $\exists x_1 \in S$, such that $f(x_1) = \max_{x \in S} f(x)$.

This means that f is bounded on S ($f(x_0) \leq f(x) \leq f(x_1)$) and attains its bounds. The maximum and minimum are attained in this setting.

Example

1) Let $f: \mathbb{R} \rightarrow \mathbb{R}$ is defined by $f(x) = x$. Let $S = [0, 1]$.

It is clear that the maximum is attained at 1 and $\max_{x \in S} f(x) = f(1) = 1$.

S It is indeed a closed and bounded set. The conditions of the theorem are satisfied.

2) On the other hand, if we study the same function but this time on $S =]0, 1[$.

We also have $\max_{x \in S} f(x) = 1$. However, the maximum is not attained : $\nexists x_0 \in S$ such that $f(x_0) = \max_{x \in S} f(x)$.

Here, the conditions of the theorem are not satisfied: S is not a closed and bounded set.

Remark: To ensure uniqueness, one must use properties specific to the objective function, such as strict convexity, for example.