

# Optimization: Basic Concepts

## Optimization

Centro de Investigación en Matemáticas A.C.

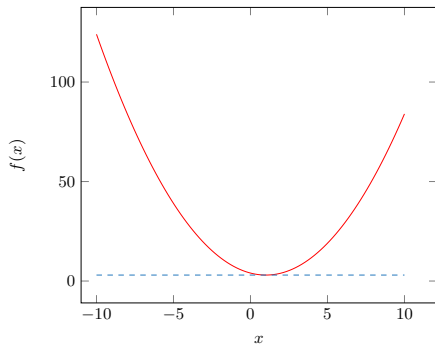


# What is optimization? I

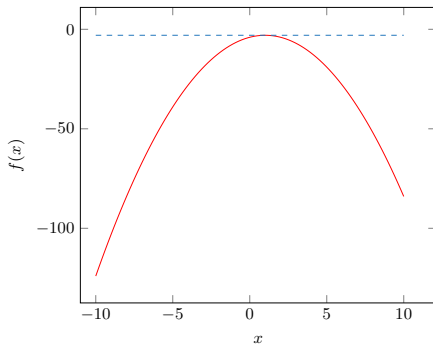
- Roughly speaking, it refers to the task of finding the best solution to a given function under some constraints
- Optimization is part of almost everything that we do:
  - Personnel schedules,
  - Routing planning,
  - Shopping,
  - Etc...



# What is optimization? II



Minimization



Maximization



# Formal Definition

Formally...

$$\begin{aligned} \min \quad & f_i(\mathbf{x}), \quad i = \{1, \dots, m\} \\ \text{Subject to: } & g_j(\mathbf{x}) \leq 0, \quad j = \{1, \dots, p\} \\ & h_k(\mathbf{x}) = 0, \quad k = \{1, \dots, q\} \end{aligned}$$

where:

- $\mathbf{x}$  is an  $n$ -dimensional vector of decision variables
- $f_i(\mathbf{x})$  is called objective function
- $g_j(\mathbf{x})$  and  $h_k(\mathbf{x})$  are called inequality and equality constraints, respectively is an equality constraint

# The objective function

- It expresses the main goal of the problem which is either to be minimized or maximized
- Problems can have:
  - no objective function,
  - a single objective function,
  - multiple objective functions



# Minimization or maximization?

An optimization problem can be written as a minimization problem or as a maximization problem. These two problems are easily converted to the other form:

$$\begin{aligned}\min f(\mathbf{x}) &\iff \max -f(\mathbf{x}) \\ \max f(\mathbf{x}) &\iff \min -f(\mathbf{x})\end{aligned}$$



# Decision Variables

- Decision variables represent the parameters that need to be determined to solve the problem
- They control the value of the objective function
- Defining the set of variables to a problem is one of the most difficult and crucial steps when formulating an optimization problem



# Constraints

- Constraints narrow the admissible values of the decision variables
- They can be
  - implicit:

$$0 \leq x_i \leq 10$$

- explicit:

$$x_1 + x_2 - x_3 = 0$$

$$x_1 - x_3 \leq 0$$





# Feasible Solution and Feasible Region I

- A solution that meets the constraints is known as **feasible solution**.

$[5, 5, 5]$

$[9, 5, 5]$

$[9, 0, 9]$

$[3, 1, 4]$



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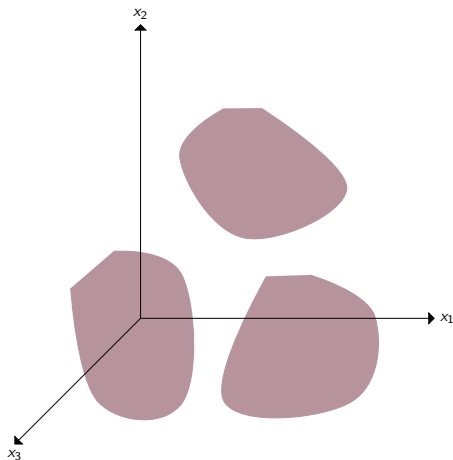


[3, 1, 4]



# Feasible Solution and Feasible Region II

- The set of all feasible solutions is known as the **feasible region**



# Local and Global Optima I

## Definition

A solution  $\mathbf{x}^*$  is called local optimum if and only if

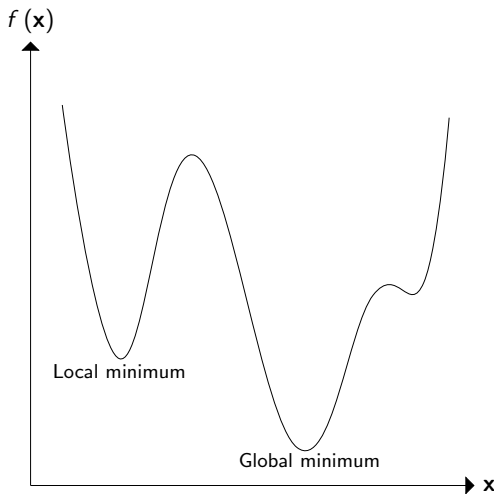
$$\forall \mathbf{x} \in \mathbf{X}_{\text{vecindad}} : f(\mathbf{x}^*) \triangleleft f(\mathbf{x})$$

## Definition

A solution  $\mathbf{x}^*$  is called global optimum if and only if  $\forall \mathbf{x} \in \mathbf{X} : f(\mathbf{x}^*) \triangleleft f(\mathbf{x})$



# Local and Global Optima II





# Classification I

Optimization problems can be categorized based on:

- their constraints
  - Constrained optimization problems
  - Unconstrained optimization problems
- their functions;
  - Separable optimization problems
  - Non-separable optimization problems
- the sort of solution;
  - Continuous optimization problems
  - Combinatorial optimization problems



# Classification II

- their decision variables;
  - Static optimization problems
  - Dynamic optimization problems
- the values on decision variables;
  - Integer programming
  - Real-valued optimization
  - Mixed-integer optimization



# Classification III

- the sort to equations;
  - Quadratic programming
  - Geometric programming
  - Linear programming
  - No-linear optimization
- the number of objectives
  - Single objective optimization
  - Multi-objective optimization
  - Many objective optimization



# Black-box Optimization I

- It refers to a problem setup in which an optimization algorithm is supposed to optimize an objective function in a so-called black-box fashion
- Problems of this type regularly appear in practice:
  - When optimizing parameters of an unknown model
  - When modeling is too complex



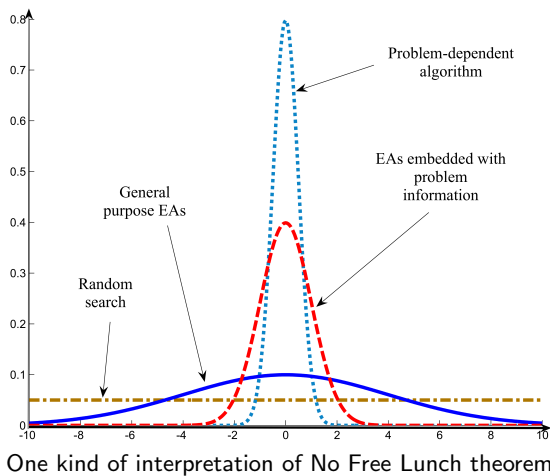
# Black-box Optimization II

There are a number of algorithms for black-box optimization

- Random Search,
- Pattern Search,
- Nelder-Mead Simplex,
- Evolutionary Algorithms,
- Particle Swarm Optimization,
- Simulating Annealing Search,
- Bat Search,
- Bee Search,
- ...



# No Free Lunch Theorem



Retrieved from Xinjie Yu and Mitsuo Gen (2010). Introduction to Evolutionary Algorithms. Springer-Verlag.

# The Exploration–Exploitation Dilemma I

- **Exploration** refers to exploring unknown regions with the aim of gaining new knowledge
- **Exploitation** refers to delving in what it is known with the aim of getting something close to what it is expected

**Should we explore or exploit?**



# The Exploration–Exploitation Dilemma II

- An “intelligent search” requires the proper balance of exploration and exploitation
- The proper balance of exploration and exploitation depends on how regular our environment is
- If our environment is rapidly changing, then our knowledge quickly becomes obsolete and we cannot rely as much on exploitation
- However, if our environment is highly consistent, then our knowledge is dependable and it may not make sense to try very many new ideas





Thanks for your attention!

Questions?

