

# Fundamentals of optimization Part I

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# Optimization

What is it?

Why do we care?

When do we use it?





# Optimization

#### **Decision variables**

$$\min c(x)$$
$$x \in F$$

One objective (objective function)

Constraints (feasible region)





# Optimization

#### One objective only?





Safety



# Fair traffic assignment

Environmentally friendly

'Low' cost















# **Optimization problem**

 $\min c(x)$  $x \in F$ 







# Local and global optima





## What do we need?

- We know how to analitically find local optima for 'simple' optimization problems
- We know how to analitically find the global optimum for some 'very simple' optimization problems
- We do not know for all the others algorithms needed



### Convex problem

 $\min_{\substack{x \in F}} c(x)$ 

Fc(x) Convex set Convex function on F



#### Convex problem





### Convex problem

#### **Property:**

For a convex problem, any local optimum is a global optimum

Effective and efficient algorithms





## Convex problem

$$\min \varphi(x) g_i(x) \ge 0 \quad (i = 1, ..., q) h_j(x) = 0 \quad (j = 1, ..., p)$$

#### Theorem:

If  $\varphi(x)$  is convex,  $g_i(x)$  are concave,  $h_j(x)$  are linear, the optimization problem is a convex problem



# Linear programming

 $\min c'x$ Ax = b $x \ge 0$ 

#### Special case of convex problem









# Linear programming

 $\min c'x \\ Ax = b$ 

 $x \ge 0$ 

#### LP problems can be solved with efficient algorithms: Simplex method Interior point methods (Khachiyan, Karmarkar)



# **Computational complexity** or simply **complexity** of an algorithm is the amount of resources required to run it

Resources: time and memory



Worst-case complexity of an algorithm

 $n \rightarrow f(n)$ *n*: size of the input

f(n) = max f(I) over all instances I of size n

Problem: Sorting Instance I: 6, 35, 7, 15, 27, 12, 18 Algorithm: Bubblesort Size: 7 f(7) = max f(I) over all instances I of size 7



f(n) very difficult to obtain

Asymptotic behavior (when *n* tends to the infinity) The complexity is expressed by using big O notation Complexity of Bubblesort:  $O(n^2)$ 



The complexity of a problem is the infimum of the complexities of the algorithms that solve the problem, including unknown algorithms.

Complexity of problem Sorting: O(n log n) (thanks to Heapsort)



Why do we care about computational complexity?

 $O(n \log n)$ 



 $O(2^{n})$ 

 $O(n^2)$ 



CPU speed \*1000

O(n)

size solved increases by 10

 $O(2^{n})$ 



George Dantzig (1914-2005) in 1947 invented the **simplex method** Worst-case complexity:  $O(2^{\frac{m}{2}-1})$ 

In practice: Rarely time required is greater than O(m log n)

Complexity of LP unknown until 1979

Leonid Khachiyan (1952-2005) in 1979 invented the ellipsoid method

Worst-case complexity:  $O(n^6)$ 

In practice:  $O(n^6)$ 







## Solution of LP

#### More and more powerful software available



(CPLEX, Gurobi) (GLPK, LP-SOLVE)





# Planning problems













**Robust optimization** 

### Mixed integer linear programming **Fixed costs** Scheduling Location min c'xAx = b $x \ge 0$ Integer **Selection** Routing





# Mixed integer linear programming

Branch-and-bound

Branch-and-cut

**Branch-and-price** 





### **Branch-and-price**





### MILP

#### More and more powerful software







### Mixed integer linear programming

**Fixed costs** 

Location

Specialized algorithms (exact and heuristic)





Routing

Scheduling