# Chapter 6: Optimization problems ENSIIE - Computational complexity theory

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# Optimization problem

## (Informal) definition

An optimization problem is a problem containing inputs (the *instances*), solutions (the *feasible solutions*), a *measure* or *weight* associating to each solution an integer. The objective is to find a solution maximizing or minimizing the measure.

# Problems associated with an optimization problem

### (Informal) definition

An optimization problem  $\Pi$  is associated with 3 problems:

- a decision problem  $\Pi_D$ : does there exist a feasible solution with weight
  - lower than K?, if the objective is to minimize the measure,
  - ullet greater than K?, if the objective is to maximize the measure ;
- an evaluation problem  $Pi_E$ : find the weight of an optimal solution;
- a construction problem  $\Pi_C$ : find an optimal solution and its weight.

## PO and NPO

#### Definition

PO and NPO are the equivalent classes of P and NP for the optimization problems: the optimization problems that can be solved in deterministic or non deterministic polynomial time.

Particularly,

- $\Pi \in PO \Rightarrow \Pi_D \in P$
- $\Pi \in NPO \Rightarrow \Pi_D \in NP$

# Optimization problem

## Definition: optimization problem

An optimization problem  $\Pi$  is a quadruplet  $(\mathcal{I}, \mathcal{S}, \mathcal{M}, \mathcal{O})$  such that:

- $\mathcal{I}$  is the set of d'instances of  $\Pi$ ;
- S is a function associating to  $x \in \mathcal{I}$  a set of feasible solutions of x;
- $\mathcal{M}$  is a *measure* function associating to  $x \in \mathcal{I}$  and  $y \in \mathcal{S}(x)$  an integer ;
- O is the objectif with value min or max that specify whether we want to find a solution with minimum of maximum measure.

 $\mathcal{I}$  and  $\mathcal{S}(x)$  may contain any mathematical objects.

# Optimal solution

#### Definition

Let  $\Pi = (\mathcal{I}, \mathcal{S}, \mathcal{M}, \mathcal{O})$  be an optimization problem and  $x \in \mathcal{I}$ . We call *optimal solution* a solution  $y^* \in \mathcal{S}(x)$  such that:

$$\mathcal{M}(x, y^*) = \min_{y \in \mathcal{S}(x)} \mathcal{M}(x, y) \text{ si } \mathcal{O} = \min$$

$$\mathcal{M}(x, y^*) = \max_{y \in \mathcal{S}(x)} \mathcal{M}(x, y) \text{ si } \mathcal{O} = \max$$

We denote the set of optimal solutions of x by  $S^*(x)$  and the value  $\mathcal{M}(x, y^*)$  of the optimal solutions by  $\mathcal{M}^*(x)$ .

## Size of an instance or of a solution

#### Definition

The size of an instance or of a feasible solution is the number of bits used to encode it.

We usually denote the size of the input x by |x| or n and the size of a feasible solution y by |y|.

# Problems associated with an optimization problem

#### (Informal) definition

An optimization problem  $\Pi = (\mathcal{I}, \mathcal{S}, \mathcal{M}, \mathcal{O})$  is associated with 3 problems:

- a decision problem  $\Pi_D$ : let  $x \in \mathcal{I}$  and an integer K, determine if  $\mathcal{M}^*(x) \leq K$  if  $\mathcal{O} = \min$ , or if  $\mathcal{M}^*(x) \geq K$  if  $\mathcal{O} = \max$ .
- an evaluation problem  $Pi_E$ : let  $x \in \mathcal{I}$ , compute  $\mathcal{M}^*(x)$ ;
- a construction problem  $\Pi_C$ : let  $x \in \mathcal{I}$ , compute an optimal solution  $y \in \mathcal{S}^*(x)$  and  $\mathcal{M}^*(x)$ .

## Some results

## **Theorem**

Let  $\Pi$  be an optimization problem:

$$\Pi_D \preccurlyeq_{\mathcal{T}} \Pi_E \preccurlyeq_{\mathcal{T}} \Pi_C$$

## **NPO**

#### **Definition**

Let  $\Pi=(\mathcal{I},\mathcal{S},\mathcal{M},\mathcal{O})$  be an optimization problem,  $\Pi$  belong to the NPO class if

- let x be a binary number, we can check in polynomial time if x encodes an instance of I;
- there exists a polynom q such that for every  $x \in \mathcal{I}$ 
  - for every  $y \in \mathcal{S}(x)$ ,  $|y| \leq q(|x|)$ ,
  - for every binary number y such that  $|y| \le q(|x|)$ , we can check in polynomial time if y encodes a feasible solution of x;
- for every  $x \in \mathcal{I}$  and every  $y \in \mathcal{S}(x)$ , we can compute  $\mathcal{M}(x,y)$  in polynomial time.

## Definition

A problem  $\Pi \in \mathsf{NPO}$  belongs to the class PO if we can solve  $\Pi_{\mathcal{C}}$  in polynomial time.

By definition  $PO \subseteq NPO$ .

# PO, NPO, P and NP

#### Theorem

$$\Pi \in PO \Rightarrow \Pi_D \in P$$

$$\Pi \in \textit{NPO} \Rightarrow \Pi_{\textit{D}} \in \textit{NP}$$

$$\Pi \in \mathit{NPO} \Rightarrow \Pi_E \preccurlyeq_{\mathcal{T}} \Pi_D$$

# NP-Hard optimization problem

#### Definition

Let  $\Pi_1$  be an optimization problem,  $\Pi_1$  is NP-Hard, for every decision problem  $\Pi_2 \in \text{NP}$ ,  $\Pi_2 \preccurlyeq_{\mathcal{T}} \Pi_1$ .

#### Theorem

Let  $\Pi \in NPO$ , if  $\Pi_D$  is NP-Hard, then  $\Pi$  is NP-Hard.

#### Theorem

$$PO = NPO \Rightarrow P = NP$$

# A hard proof

#### **Theorem**

Let  $\Pi \in \mathsf{NPO}$ , if  $\Pi_D$  is  $\mathsf{NP\text{-}Complete}$ , then  $\Pi_C \preccurlyeq_{\mathcal{T}} \Pi_D$ .