

## Optimization Methods: Theory and Applications Introduction

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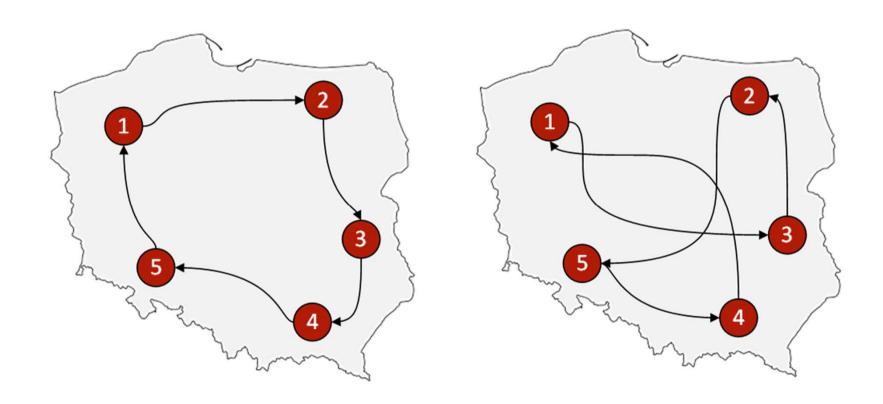
### **Travelling Salesman Problem**

- Find the shortest route
- Visit each city exactly once





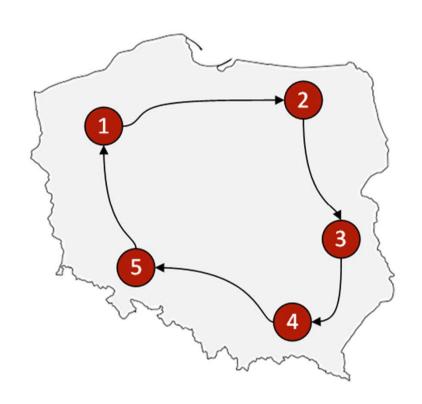
### **Travelling Salesman Problem**

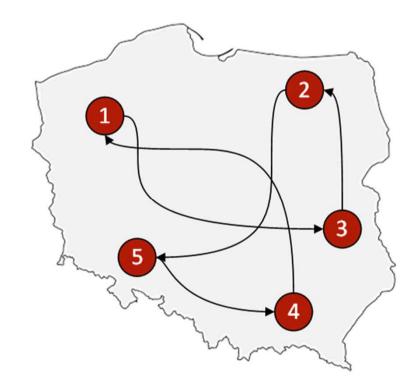


 $\frac{n!}{2n}$  available routes (for n=5 there is 12 of tchem)



### **Travelling Salesman Problem**





[1, 2, 3, 4, 5]

[1, 3, 2, 5, 4]



### Optimization – what do you want?

# As good solution as possible...



### **Optimization**

- Different problems have different features
- There are optimizer is optimal for every problem
- What do we want: use the features of the optimized problem (to our favour)



## **Optimization – what is it?**Traveling Salesman Problem

• Single solution:

$$\vec{x} = [x_1, x_2, x_3, x_4, x_5]$$

$$\forall_{i \in \{1, \dots, 5\}} x_i \in D_{x_i} = \{1, \dots, 5\}$$

The set of all solutions:

$$D_f = D_{x_1} \times D_{x_2} \times D_{x_3} \times D_{x_4} \times D_{x_5}$$

The set of feasible solutions:

$$D_{\vec{x}} = \left\{ \vec{x} \colon \forall_{i \in \{1, \dots, 5\}} \ \forall_{j \in \{1, \dots, 5\} - \{i\}} \ x_i \in D_{x_i} \land x_j \in D_{x_j} \land x_i \neq x_j \right\}$$



## **Optimization – what is it?**Traveling Salesman Problem

Optimized function:

$$f: D_f \to R$$

$$f(\vec{x}) = \left(\sum_{i=1}^4 \delta(x_i, x_{i+1})\right) + \delta(x_5, x_1)$$

where:  $\delta(x, y)$  – the distance between cities x and y

• Objective: find solution  $\vec{x}^*$ :

$$\vec{x}^* = \arg\min_{\vec{x} \in D_{\vec{x}} \subseteq D_f} f(\vec{x})$$



# **Optimization – what is it?**Various solution spaces

#### Discrete problems

$$D_f = Z_1 \times \cdots \times Z_n$$

Example:

$$D_f = \{1, 2, 3\} \times \{4, 5\} \times \{9, 10, 12, 14\}$$
  
 $\vec{x} = [2, 5, 12]$ 

Binary problems (the subset of discrete problems)

$$D_f = \{0, 1\}^n$$

Example:

$$D_f = \{0, 1\} \times \{0, 1\} \times \{0, 1\} = \{0, 1\}^3$$
$$\vec{x} = [0, 1, 1]$$

Continuous problems

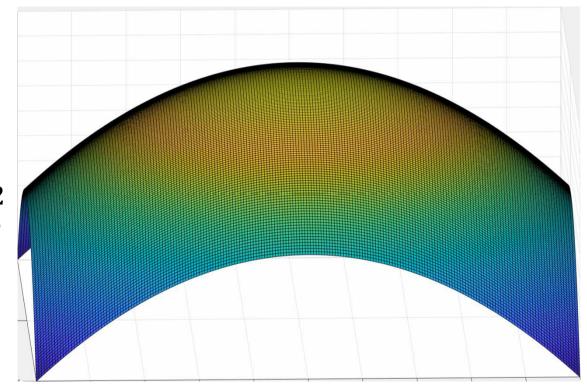
$$D_f \subseteq R^n$$

Example:

$$D_f = [-5, 5] \times [-5, 5] \times [-5, 5] = [-5, 5]^3$$
  
 $\vec{x} = [3.14, 2.56, 1.98]$ 

### **Problem nature**

- *Hill* problem
- $\vec{x} = [x_1, x_2]$
- $f: D_f = R^2 \to R$
- $f(\vec{x}) = -x_1^2 x_2^2$

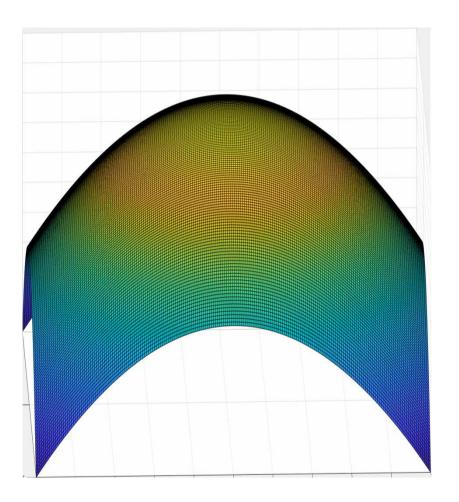


• 
$$\vec{x}^* = \underset{x \in D_{\vec{x}} = [-5,5]^2 \subseteq D_f}{\operatorname{arg max}} f(\vec{x})$$



### **Problem nature**

- The *Hill* problem features
  - Different than in the TSP case?
  - What makes the difference?
  - Why?
- How would you solve it by hand
  - TSP?
  - The *Hill* problem?
- What is the difference between the TSP and the Hill problem solutions?





#### **Problem nature**

- Travelling Salesman Problem (TSP) combinatorial in nature
  - We want to exchange solution fragments
  - For instacne the "good" city sequences
- Problem Hill topological in nature
  - We want to search for the better solution in the neighbourhood of the best solutions found that far
  - We shift "slightly left", or "slightly right"
- Different problem nature -> Different tools



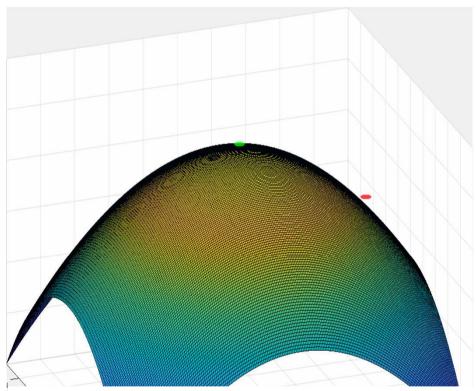
### Random Search

- Optimization how to make it in the simplest way?
- Randomly generate and remember the bestfound solution
- That's it!



### Random Search

- RS is it effective?
- What is the main drawback of RS?
- No memory
- How to handle that?
- Next lecture



The green point – the optimum

The black point – the best solution

The red point – the current solution