



# Local Search and Constraint Programming

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PART

# Introduction



# Local Search and Constraint Programming

Real-world combinatorial optimization problems

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Real-world combinatorial optimization problems:

- They are usually **large** (real-world crew scheduling applications)
- They are **not pure** (involve a heterogeneous set of side constraints)



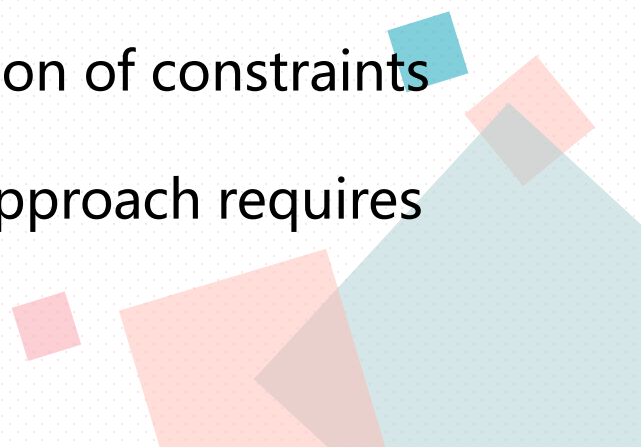
Exact approaches cannot be applied to solve real-world problems





# Local Search and Constraint Programming

Three important issues of dealing with real-world problems:

- Huge problems require large neighborhoods whose exploration can be computationally expensive
  - When dealing with problems involving many heterogeneous side constraints it is often preferable to consider them as hard constraints rather than transforming them into penalty functions.
  - Real-world applications typically lead to frequent update/addition of constraints (recall again union contract regulations), thus the algorithmic approach requires flexibility.
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# Local Search and Constraint Programming


**CP:** Modeling and solving real-world combinatorial optimization problems

- Vehicle Routing Problem
- Graph Colouring Problem
- Job Shop Scheduling
- Airline Scheduling Problem





# Local Search and Constraint Programming

- LS may use ideas from CP in order to make large neighborhoods more tractable.
  - CP may use ideas from LS to explore a set of solutions close to the greedy path in a tree search and converge more quickly towards the optimum.
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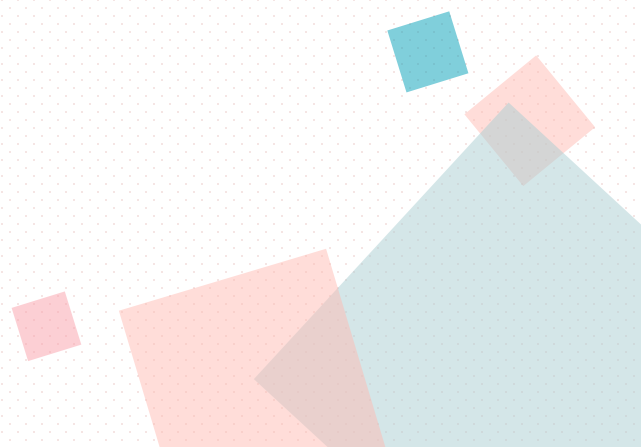
**A case of CP**





# Constraint Programming

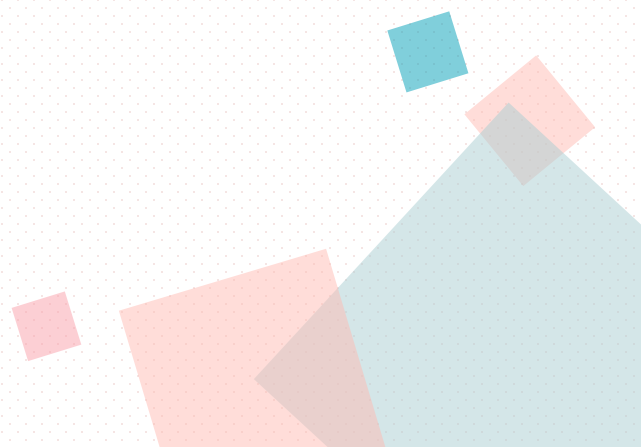
CP includes:

- Constraint Satisfaction  $\left\{ \begin{array}{l} \textit{Constraint Propagation} \\ \textit{Domain Reduction} \\ \dots \end{array} \right.$
  - Constraint Solving
- 



# Variables

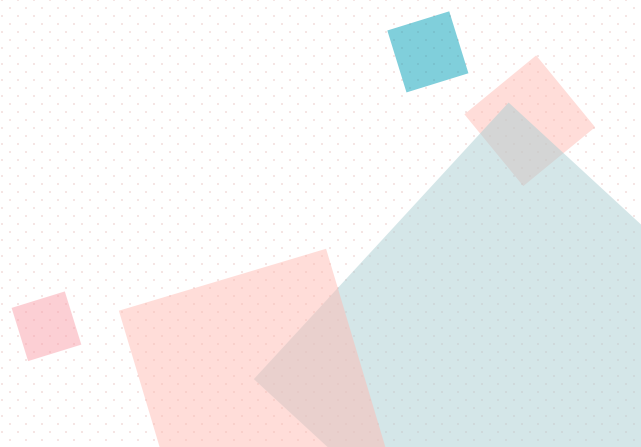
## Vehicle Routing Problem:

- $i, j \in (1, \dots, N)$  for locations (clients)
  - $k \in (1, \dots, M)$  for trucks (routes)
  - $h \in (1, \dots, 2M)$  for bins
  - $l \in (1, \dots, P)$  for types of goods
- 



# Constraints

## Vehicle Routing Problem:

- Each route is associated to a truck, and starts and ends at the depot;
  - Each client is visited exactly once, and within the time window;
  - The bins' capacity constraints are respected
- 

# CP model of Vehicle Routing Problem

$$\min \quad \text{totCost} = \sum_{k=1}^M \text{cost}_k$$

on

$$\forall k \in \{1, \dots, M\} \quad \text{cost}_k \geq 0, \\ \text{truck}_k = \text{UnaryResource}(tt, c, \text{cost}_k)$$

$$\forall h \in \{1, \dots, 2M\} \quad \text{collects}_h \in [1..P]$$

$$\forall i \in \{1, \dots, N\} \quad \text{start}_i \in [a_i..b_i], \\ \text{service}_i = \text{Activity}(\text{start}_i, d_i, i), \\ \text{visitedBy}_i \in [1..M], \\ \text{collectedIn}_i \in [1..2M]$$

subject to

$$\forall i \in \{1, \dots, N\} \quad \text{service}_i \text{ requires } \text{truck}[\text{visitedBy}_i] \quad (1)$$

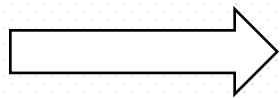
$$\forall h \in \{1, \dots, 2M\} \quad \sum_{i \mid \text{collectedIn}_i = h} q_i \leq C \quad (2)$$

$$\forall i \in \{1, \dots, N\} \quad \text{collects}[\text{collectedIn}_i] = \text{type}_i \quad (3)$$

$$\forall i \in \{1, \dots, N\} \quad \text{visitedBy}_i = \left\lceil \frac{\text{collectedIn}_i}{2} \right\rceil \quad (4)$$

# Propagation

$$\begin{aligned} & \inf(start_j) + d_j + tt_{ji} > \sup(start_i) \\ & \wedge \inf(start_i) + d_i + tt_{ij} > \sup(start_j) \\ & \wedge value(visitedBy_i) = k \end{aligned}$$



$$visitedBy_j \neq k$$

# Propagation

$$\begin{aligned} \forall i, j \in \{1, \dots, N\} \quad & next_i = j \Rightarrow \\ & succ_i = \{j\} \cup succ_j \end{aligned} \quad (11)$$

$$\begin{aligned} \forall i, j \in \{1, \dots, N\} \quad & j \in succ_i \Rightarrow \\ & visitedBy_i = visitedBy_j \end{aligned} \quad (12)$$

$$\begin{aligned} \forall k \in \{1, \dots, M\}, \forall i \in \{1, \dots, N\} \quad & first_k = i \Rightarrow \\ & visitedBy_i = k \end{aligned} \quad (13)$$

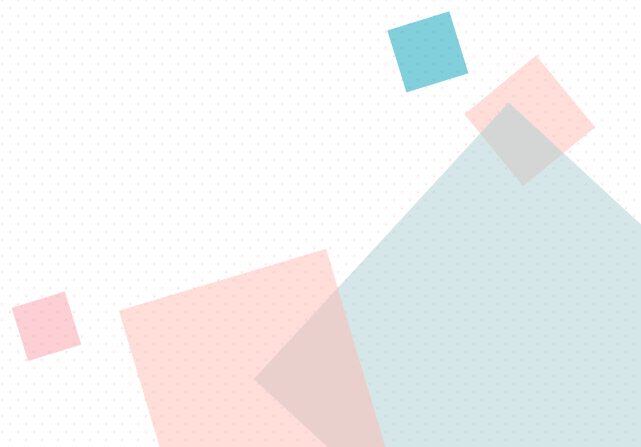


# A Case of Vehicle Routing Problem

Approximate Solution

Approximate Solution

It is **impractical** to solve the model by complete global search

- Insertion Algorithms
  - Greedy Insertion
  - Discrepancy-based Search
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## Characteristics of CP





# Characteristics of CP

CONCEPTS

- Efficient
  - Flexible
  - It can reflect only part of the characteristics of the object
  - Constraints can be established between different fields
  - Constraint has no direction (X can restrict Y, Y can restrict X)
  - Constraints are stackable
  - Constraints are often not single, a variable can have multiple constraints
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