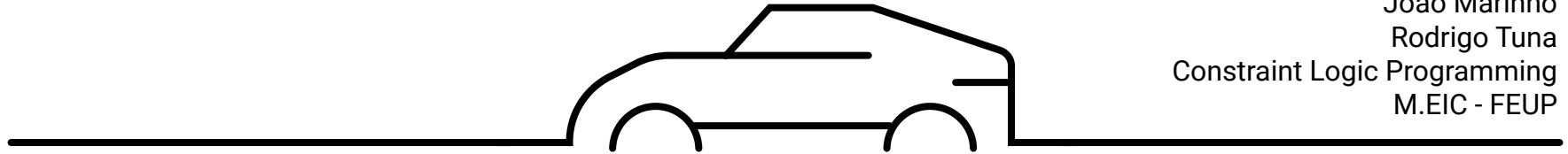


# Vehicle Routing Problem

Efficiently optimizing the path to  
deliver goods



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M.EIC - FEUP

# Problem Recap (VRP)

- The VRP is a classical optimization problem, problem that seeks to find the optimal set of routes for a fleet of vehicles to visit a set of locations (clients) minimizing the distance travelled.
- Introduced by Dantzig et al. in 1959 [1] as a generalization of the Travelling Salesman Problem.
- It is a NP hard problem and so several techniques can be used to solve it, optimization methods, heuristic methods.
- Several variants of the VRP exist 2 of which will be explored.

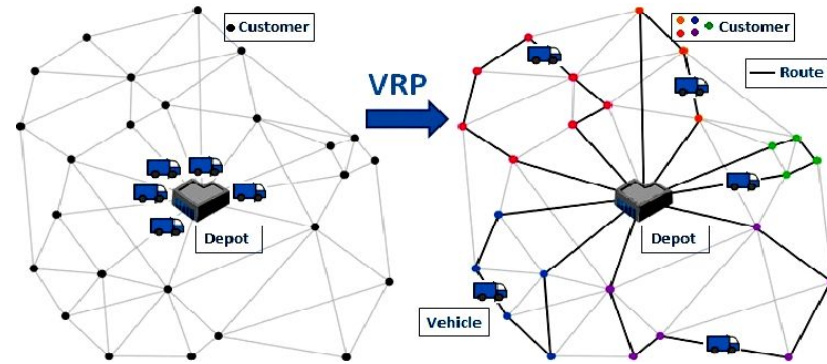


Figure 1 - Classical VRP [2]

# Problem Recap (MDVRPTW)

1

The vehicles can depart from any of the depots.

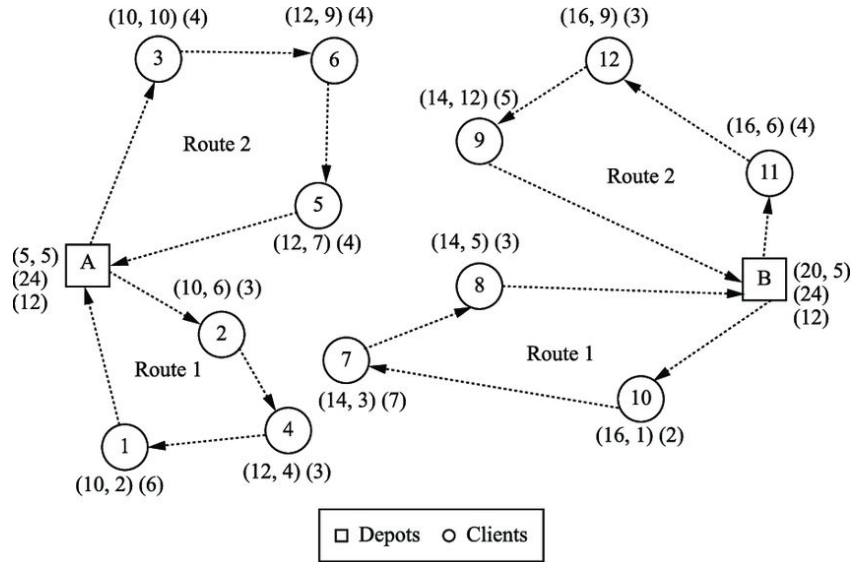


Figure 2 - Example of MDVRP [3]

2

Each client  $i$ , is associated with an interval  $[a_i, b_i]$  (time window) that specifies that a vehicle should not arrive at  $i$  after  $b_i$ .

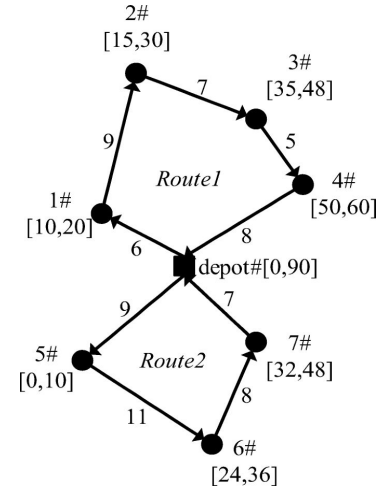


Figure 3 - Example of VRPTW[4]

# Problem Definition

Vehicle Capacity



Vehicle Route Duration



Homogeneous Fleet



Time Windows



No Traffic Congestion



**Optimal Solution**

$$\text{Cost} = \sum \text{route\_cost}_i, i \in \text{list of vehicles}$$

# CP Model - Domain Variables

Add new depot set  $D'$  containing a copy of each depot for each vehicle in the depot.

## Notation

$$Next_i \in D' \cup C, \forall i \in D' \cup C \quad (1)$$

$$A_i \in [0, T] \quad (2)$$

$$W_i \in [0, T] \quad (3)$$

## Explanation

- (1) Next node to visit
- (2) Arrival Time
- (3) Waiting Time (Only for TW)

# CP Model - Constraints

## Notation

$$Next_i \in \{i\} \cup C, \forall i \in D' \quad (4)$$

$$AllDifferent(Next_i | i \in D' \cup C) \quad (5)$$

$$A_{Next_i} = Time(i, Next_i), \forall i \in D' \quad (6)$$

$$A_{Next_i} = A_i + W_i + ServTime(i) + Time(i, Next_i), \forall i \in C \quad (7)$$

$$A_i + W_i \geq Open(i) \quad (8)$$

$$A_i \leq Close(i) \quad (9)$$

## Explanation

- (4) No inter-depot routing
- (5) Not visit node more than once
- (6) Define Arrival time for first node
- (7) Define Arrival time for next nodes
- (8,9) Time Window constraints

## CP Model - Target

**Minimize**

costs of all trips.

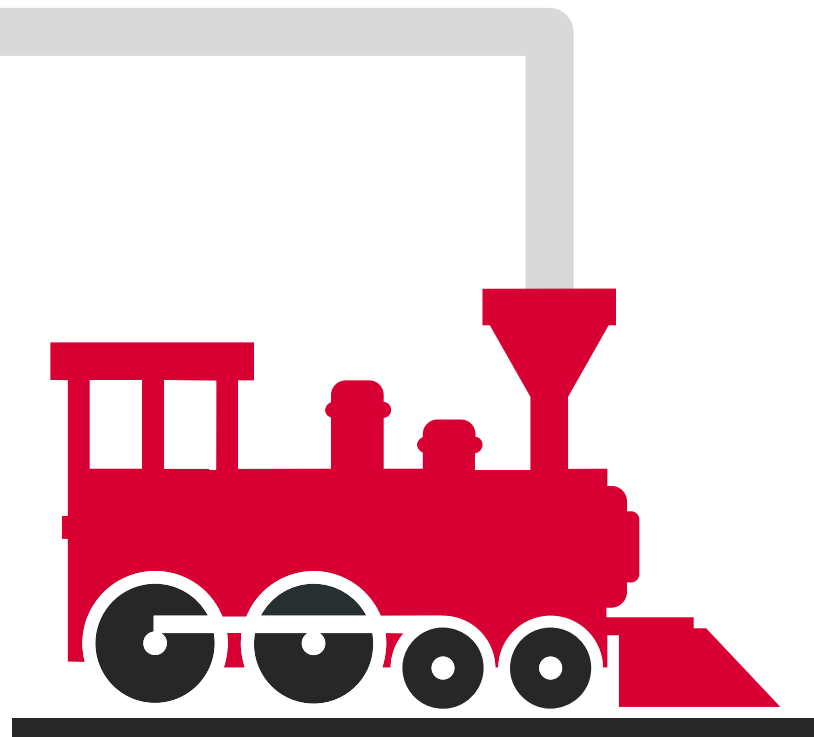
$$\sum_{i \in D' \cup C} Cost(i, Next_i)$$

# Datasets

MDVRP Dataset [5]

VRPTW Dataset [5]

MDVRPTW Dataset [5]



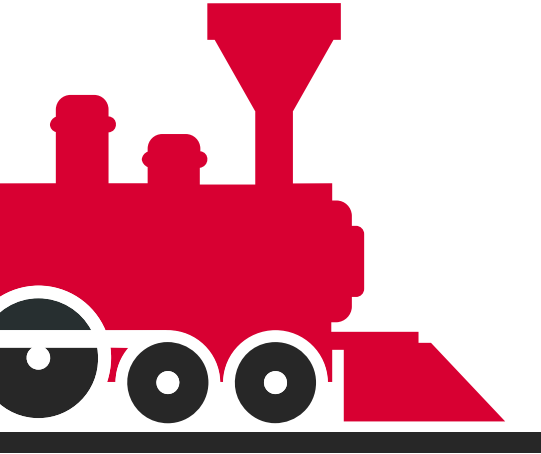


# Datasets - Structure

1. type #vehicles #customers #depots
2. vehicle\_details (max duration & load)

#customers + #depots lines:

3. id x y service\_time demand (+ discarded information)



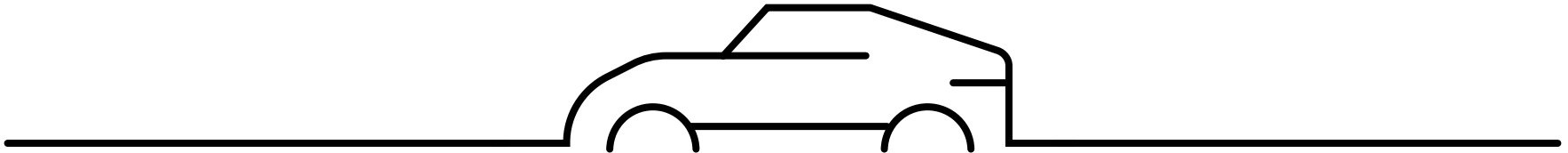
# Python Implementation

## Dimensions

- Time (Our Cost)
  - $\text{Distance}(i,j) / \text{Speed} + \text{ServiceTime}(i)$
  - 0 to maxRouteTime
  - Slack time = maxRouteTime for Time Window problems
- Capacity
  - Demand(i)
  - 0 to MaxCapacity
  - No slack

## Constraints

- Start and End Depots
- Add Time Window to each Client



# Prolog Implementation

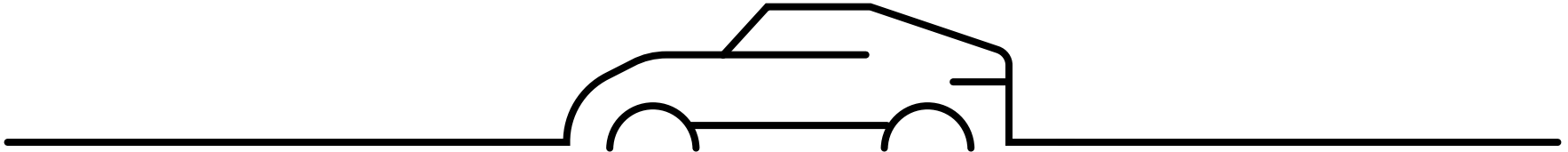
## Variables

- Routes
  - MaterializedRoutes
- LeaveTimes
- TotalRouteTimes
  - TotalTime

## Constraints

Depot, customer, demand, time window and total time

- $\text{subcircuit}(R_i), \forall i \in \text{Routes}$
- $\text{lex\_chain}(R\_Set_i), \forall i \in \text{Routes Per Depot}$
- $\text{Next\_Start\_Work} \# = \max(\text{Arrive\_Time}, \text{Open\_Time})$
- $\text{Close\_Time} \# \geq \text{Arrive\_Time}$
- $\text{Next\_Leave\_Time} \# = \text{Next\_Start\_Work} + \text{Service\_Time}$
- $\text{sum}(\text{Materialized\_Customer}, \# =, 1)$
- $\text{scalar\_product}(\text{Demands}, \text{Materialized\_Route}, \# \leq, \text{Max\_Demand})$
- $\text{Total\_Route\_Time} \# \leq \text{Max\_Route\_Time}$



# Or-Tools Experimental Setup

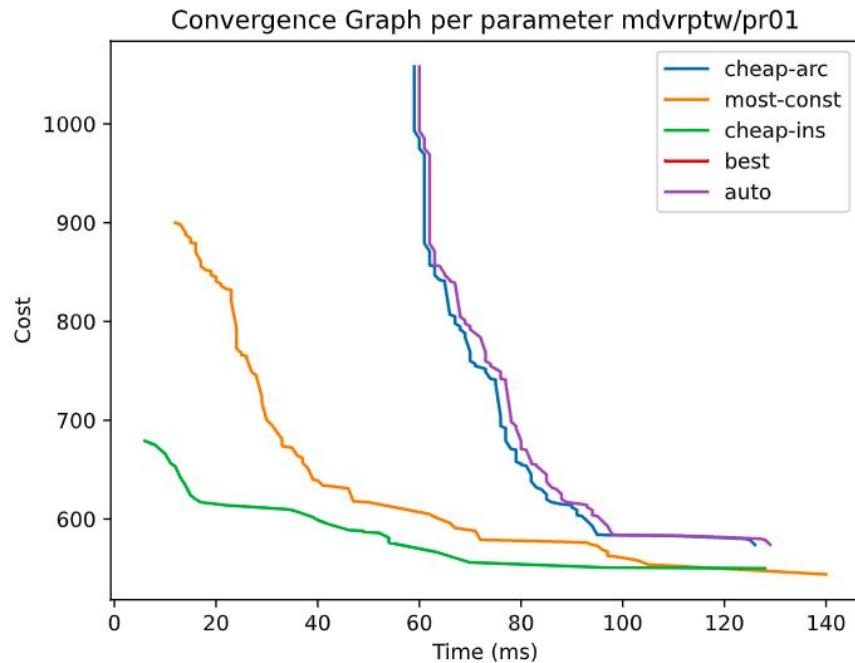
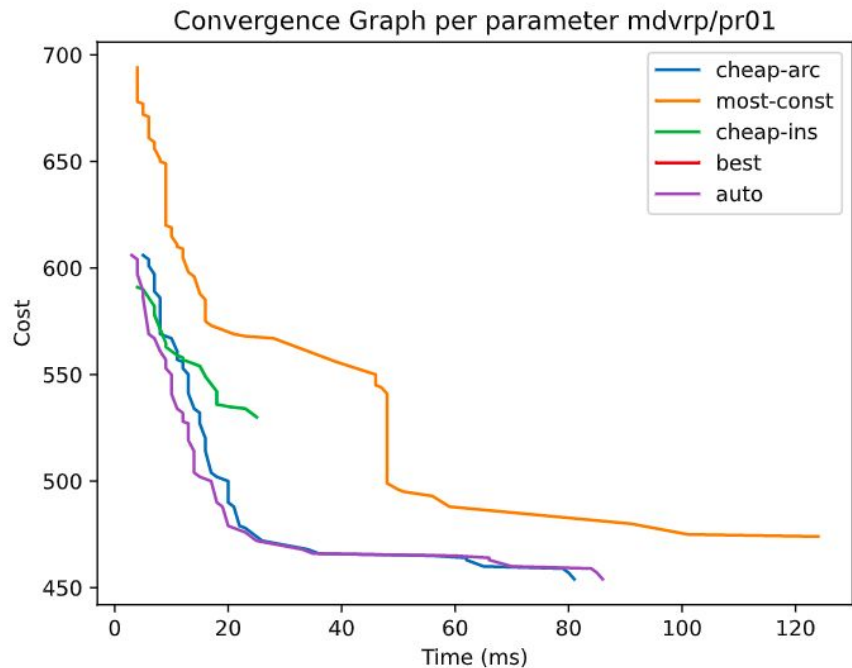
## Parametres

- AUTOMATIC
- PATH\_CHEAPEST\_ARC
- PATH\_MOST\_CONSTRAINED\_ARC
- BEST\_INSERTION
- LOCAL\_CheAPEST\_INSERTION

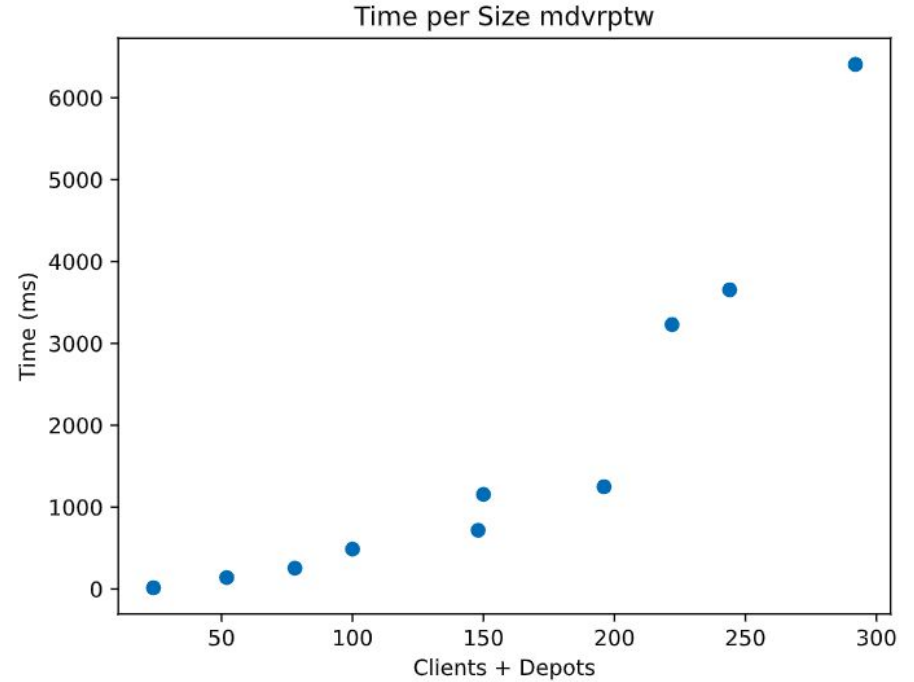
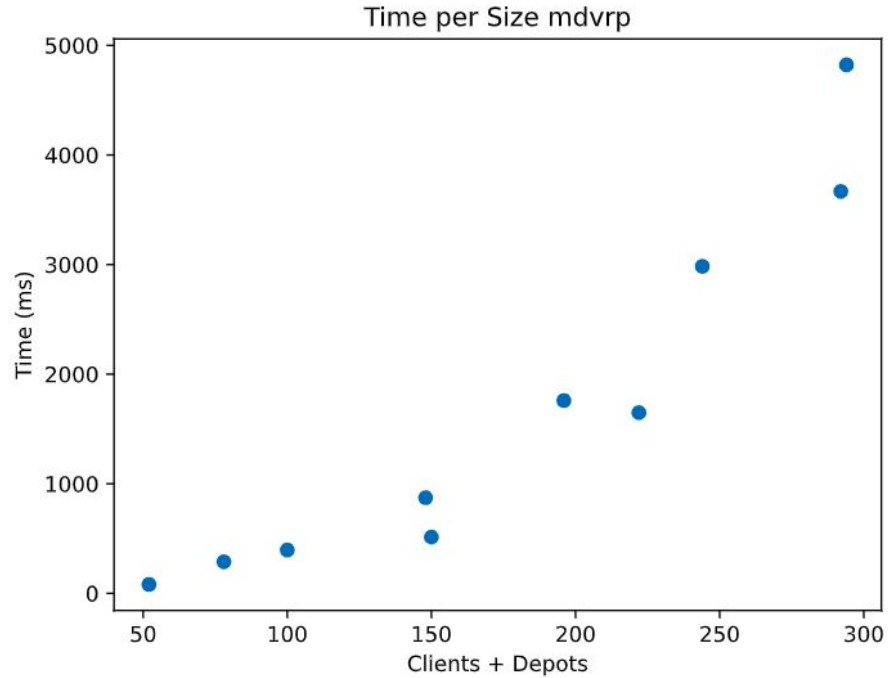
## Time

- Timeout of 20 minutes
- Collect all solutions

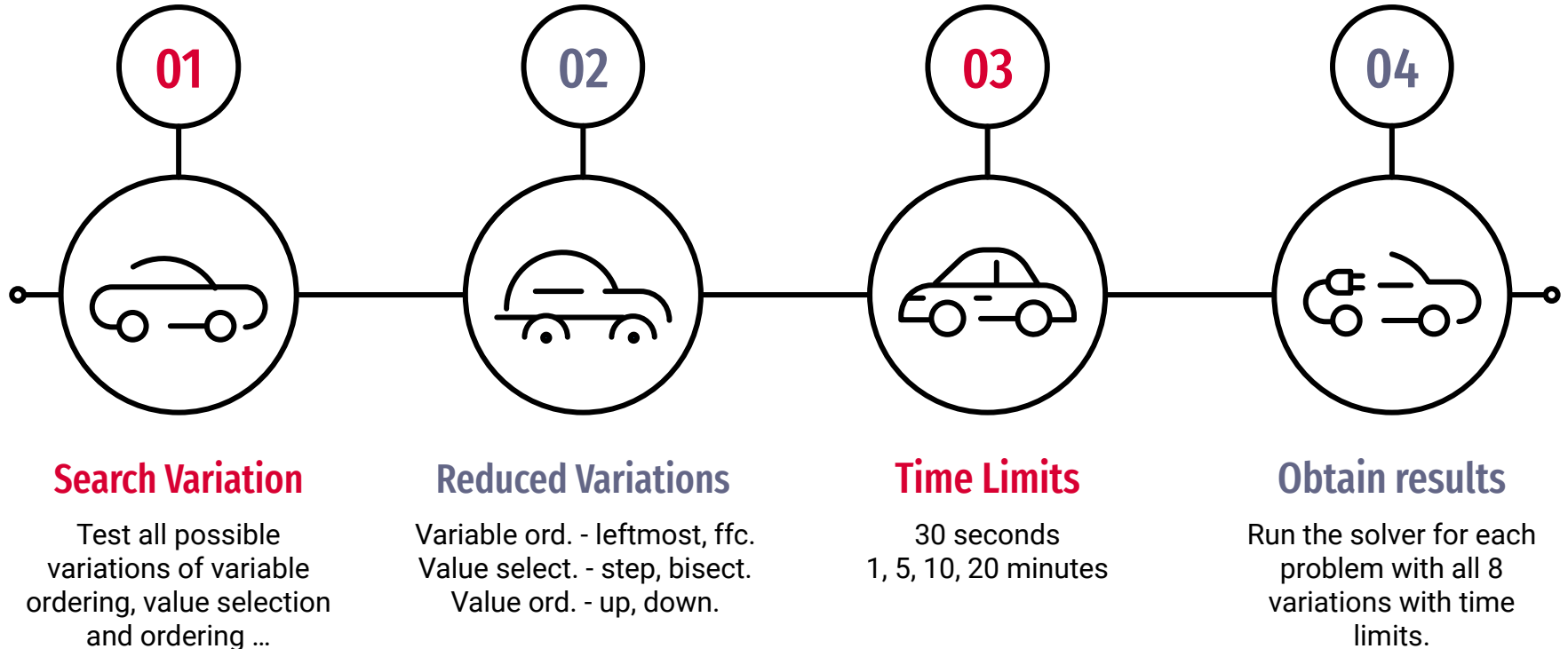
# Results - Or-Tools - Parameters



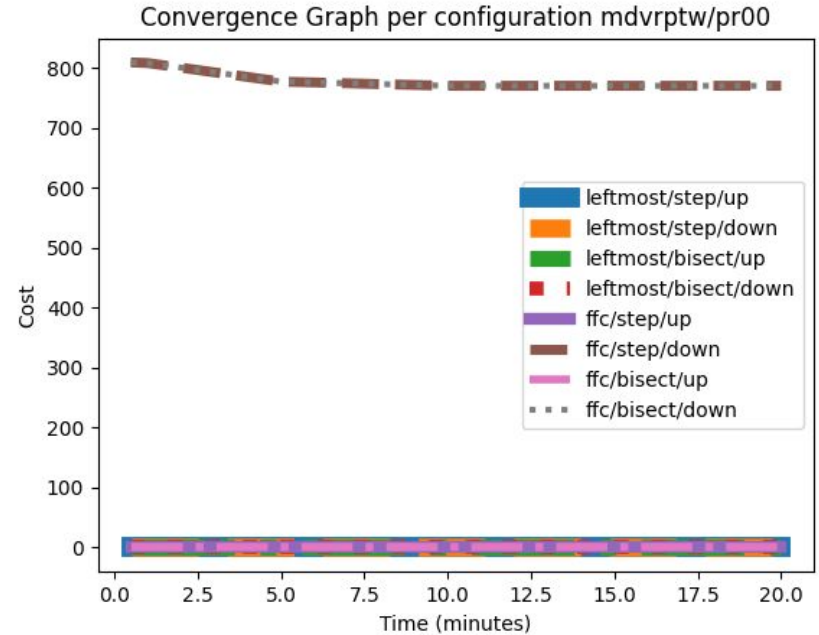
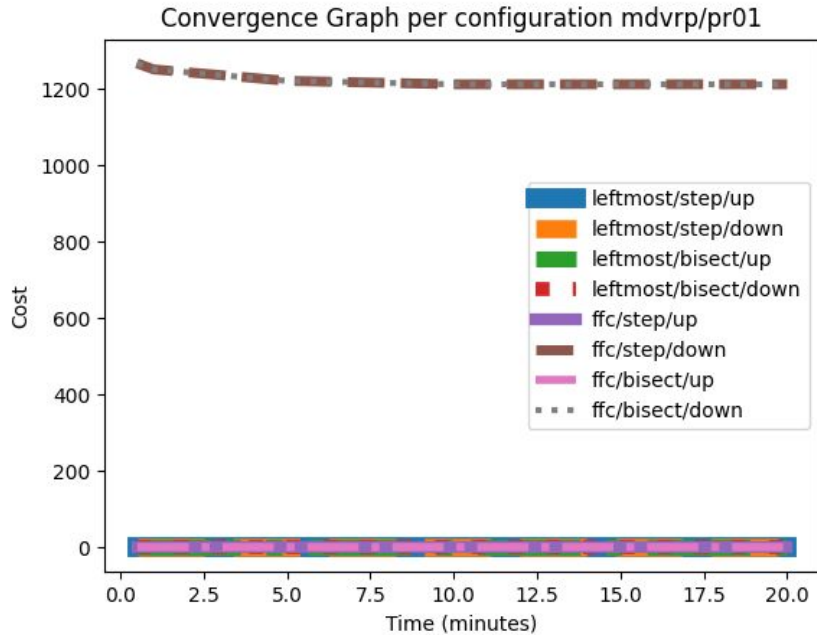
# Results - Or-Tools - Size



# Prolog Experimental Setup



# Results - Prolog





# Results - Comparison

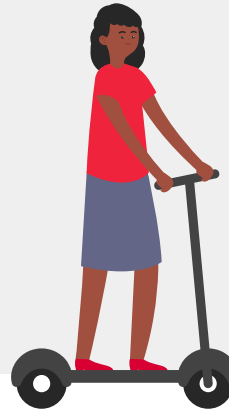
## Solution\*

Or-Tools solution is on average 2.64 times better than Prolog



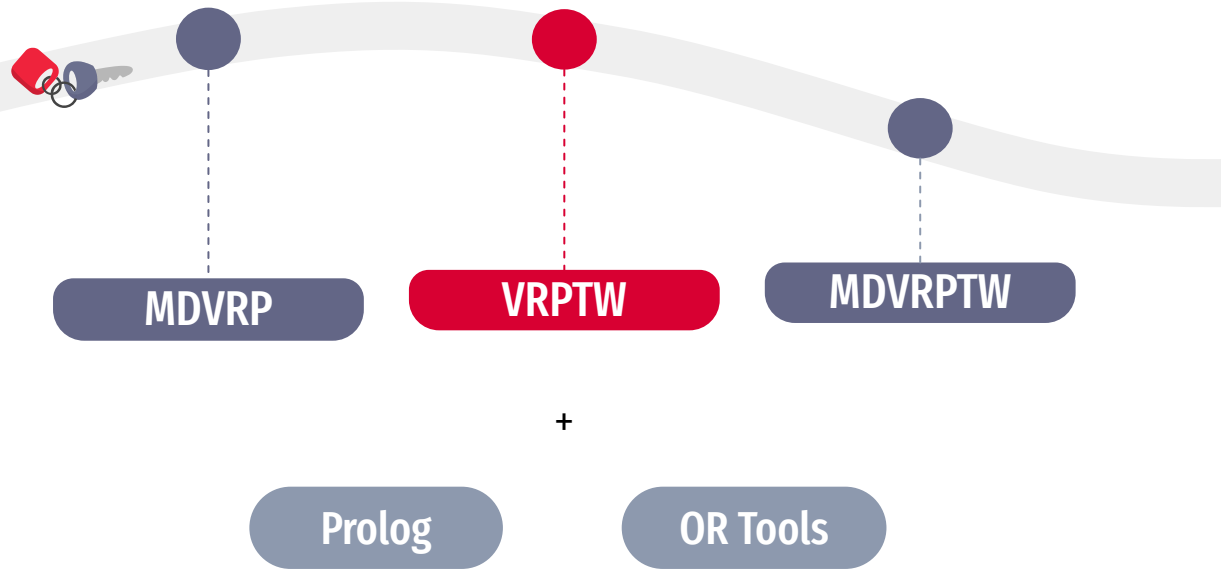
## Time\*

Or-tools achieves the best solution 23703.70 times faster than Prolog



\*For problems where Prolog found a solution (6%)

# Conclusion



# Bibliography

- [1] - G. B. Dantzig, J. H. Ramser, (1959) The Truck Dispatching Problem. Management Science 6(1):80-91.
- [2] - Gupta, Ashima & Saini, Sanjay. (2017). An Enhanced Ant Colony Optimization Algorithm for Vehicle Routing Problem with Time Windows. 267-274. 10.1109/ICoAC.2017.8441175.
- [3] - Silva Junior, Orivalde & Lopes, Luiz & Bergmann, Ulf. (2011). A Free Geographic Information System as a Tool for Multi-Depot Vehicle Routing. Brazilian Journal of Operations & Production Management. 8. 103-120. 10.4322/bjopm.2011.006.
- [4] - Feng, B., Wei, L. An improved multi-directional local search algorithm for vehicle routing problem with time windows and route balance. Appl Intell (2022).
- [5] - <http://neumann.hec.ca/chairedistributique/data>