



Aalto University
School of Science

Solving the Green Vehicle Routing Problem

Juho Andelmin

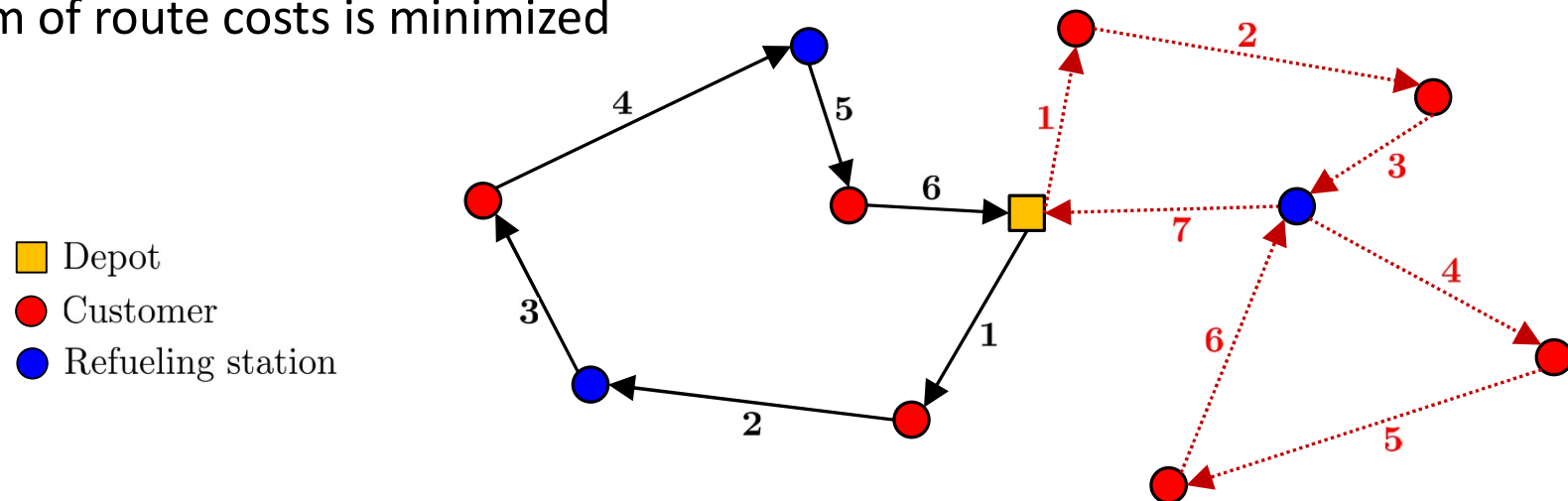
Enrico Bartolini¹

- Andelmin, J., Bartolini, E. (2017). **An Exact Algorithm for the Green Vehicle Routing Problem.** *Transportation Science*. Advance online publication. <http://doi.org/10.1287/trsc.2016.0734>
- Andelmin, J., Bartolini, E. **A Multi-Start Local Search Heuristic for the Green Vehicle Routing Problem Based on a Multigraph Reformulation.** Submitted to *Computers and Operations Research*

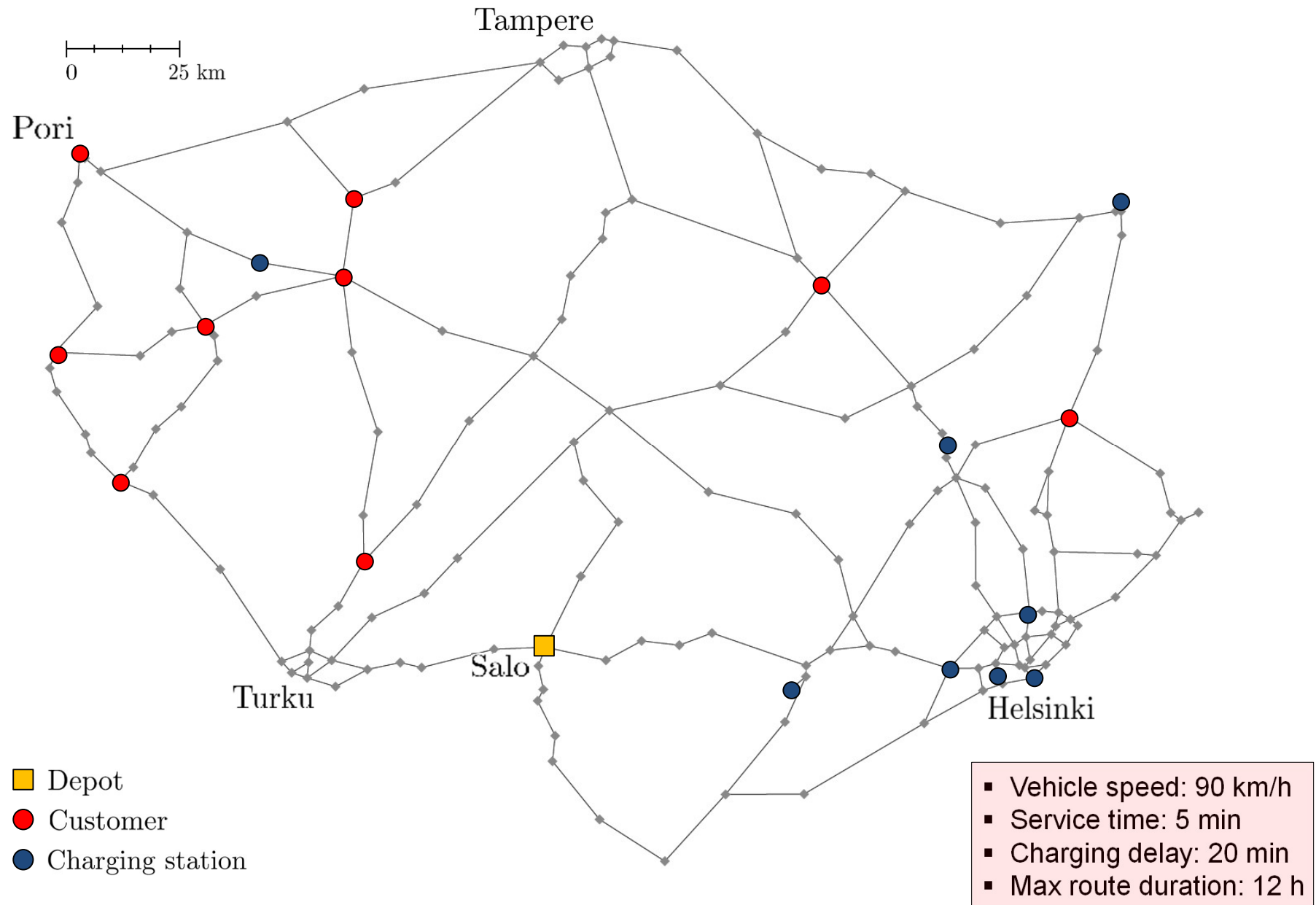
¹ RWTH Aachen University
School of Business and Economics

Green Vehicle Routing Problem (G-VRP)

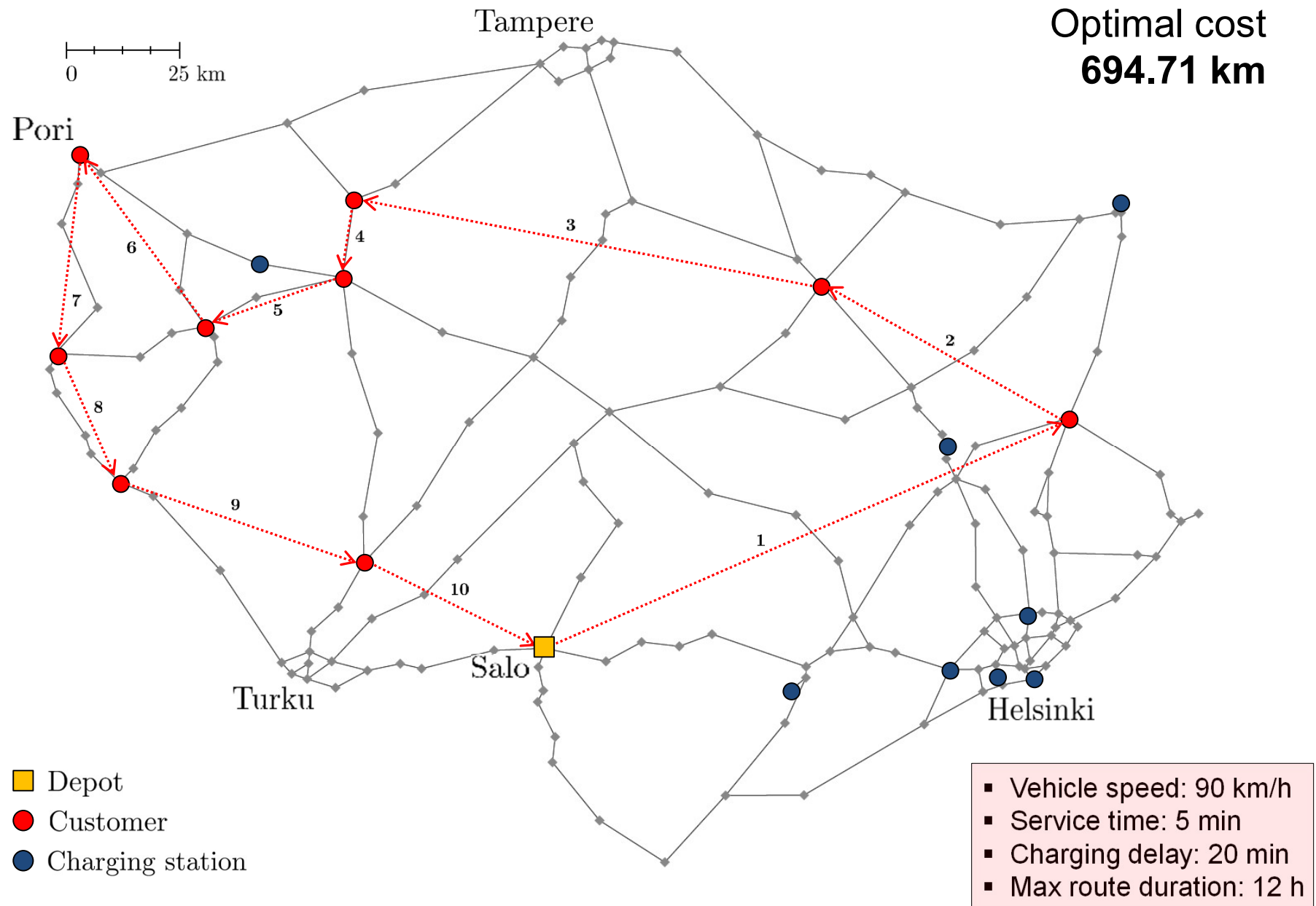
- A fleet of vehicles based at a **depot** is to serve a set of customers
 - Customers have known service times
 - Vehicles have limited fuel capacity
 - Vehicles can visit refueling stations to refuel
- Objective: Design a set of **vehicle routes** so that
 - Every customer is served
 - Duration of each route $\leq T$
 - Sum of route costs is minimized



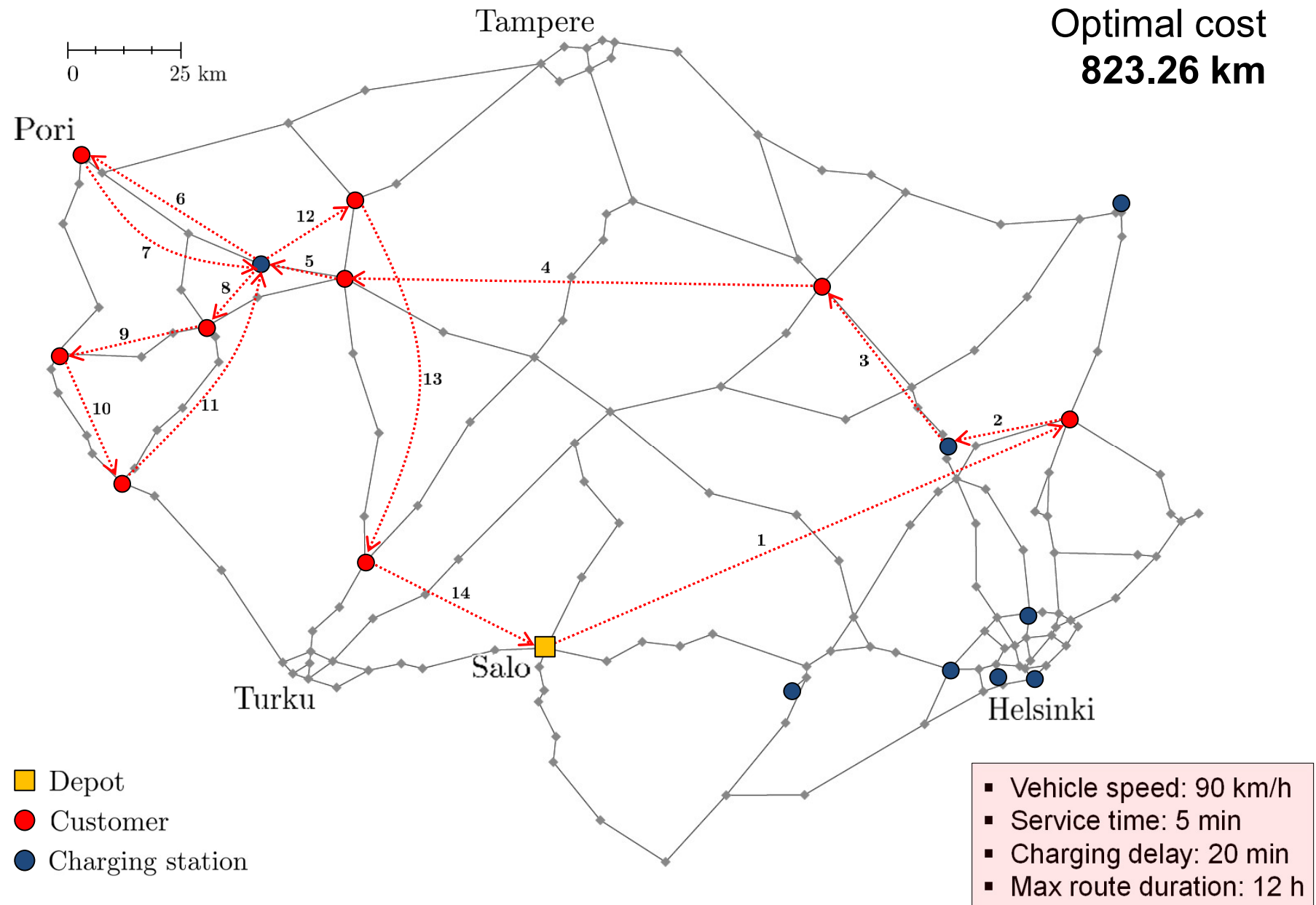
Simple example: 9 customers, electric vehicles



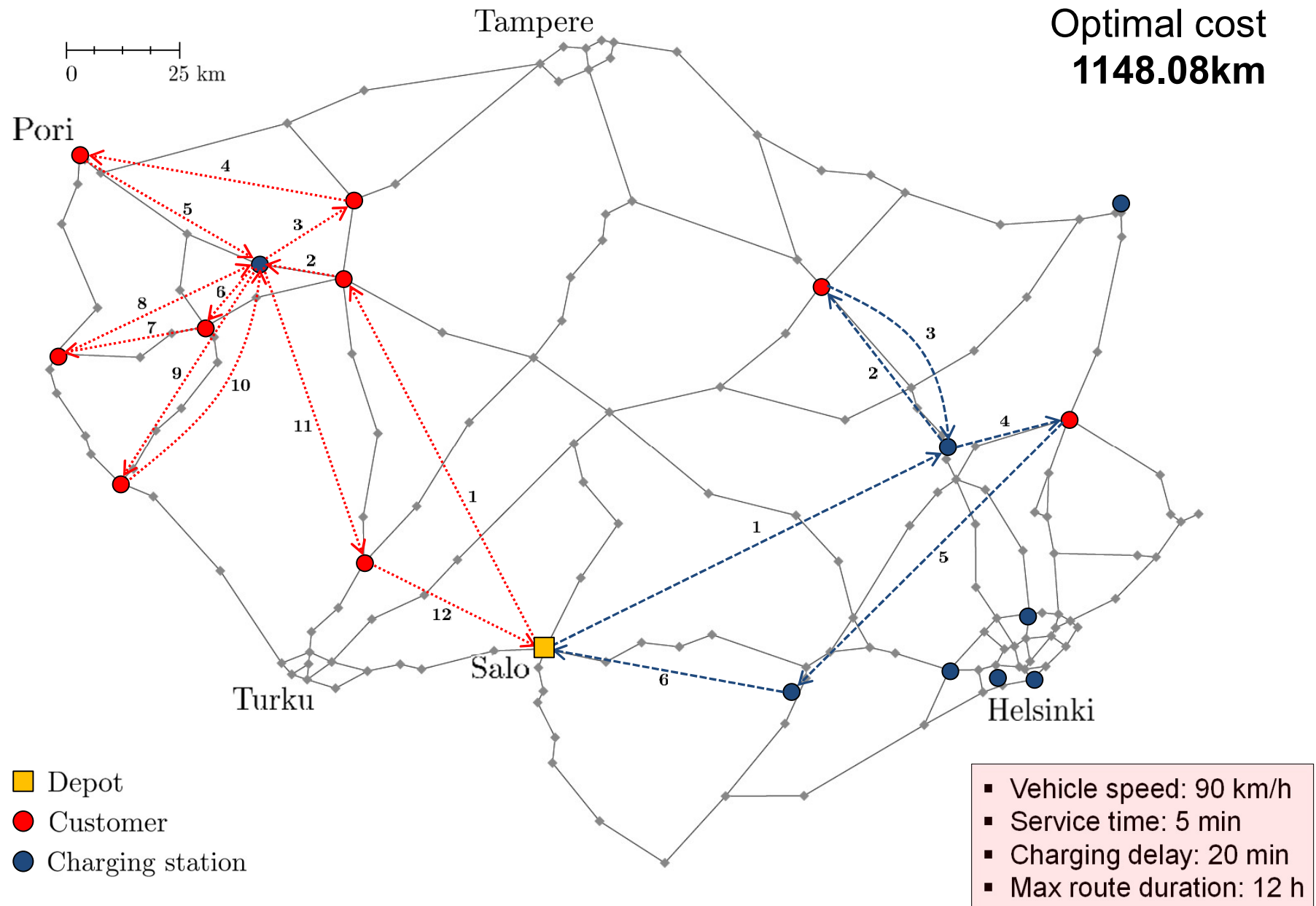
Optimal solution with driving range = ∞



Optimal solution with driving range = 200 km

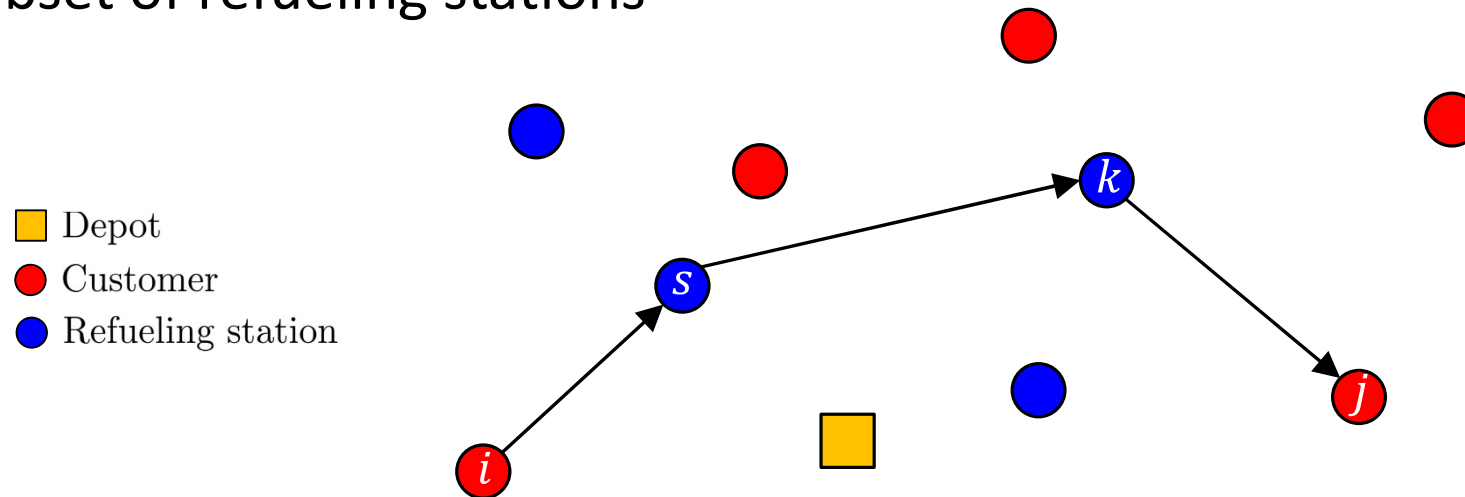


Optimal solution with driving range = 160 km

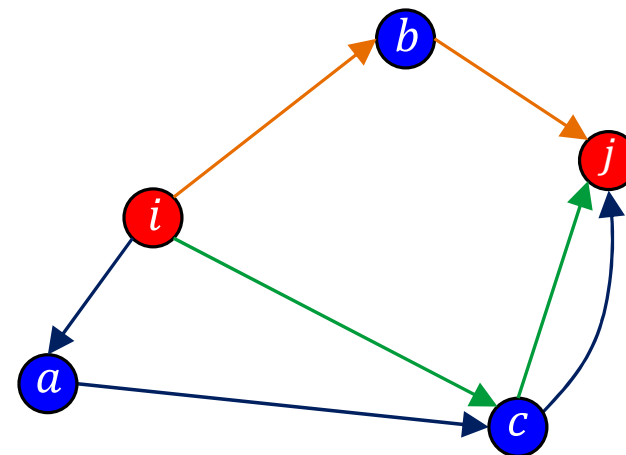


Refuel paths

- **Refuel path:** a simple path between two customers that visits a subset of refueling stations



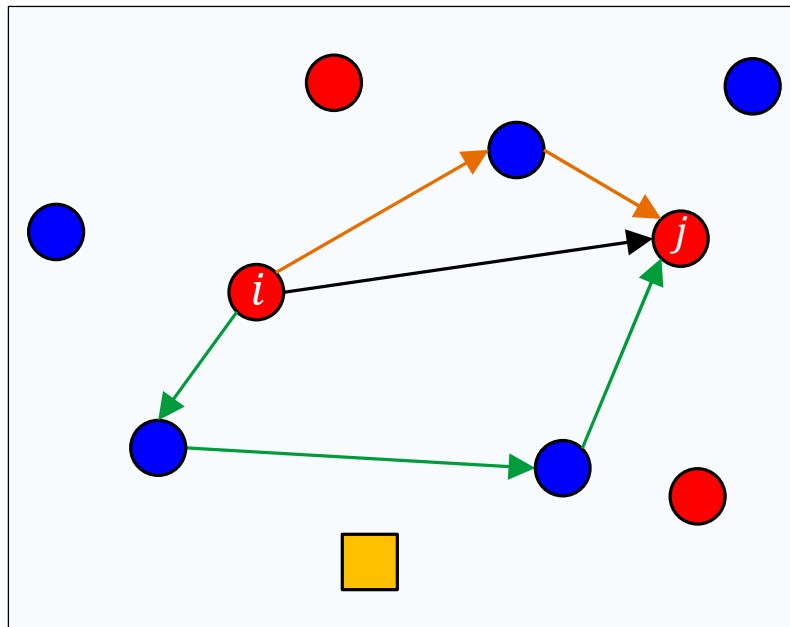
- Many refuel paths are dominated
- Example:
 - **Green path** $i \rightarrow c \rightarrow j$ is dominated by
 - **orange one** $i \rightarrow b \rightarrow j$



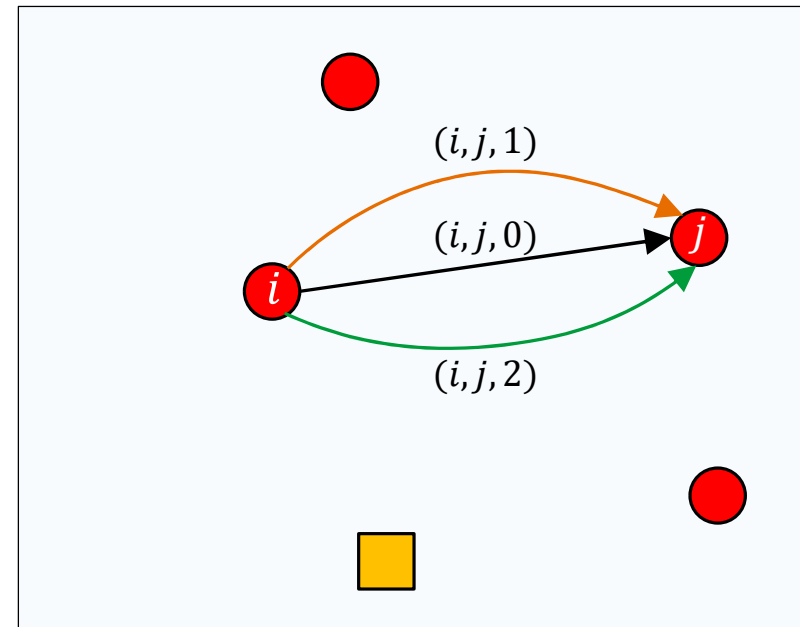
Multigraph

- We model the G-VRP on a multigraph \mathcal{G} with one arc for each non-dominated refuel path

Two refuel paths + direct arc from i to j



Three corresponding arcs in \mathcal{G}



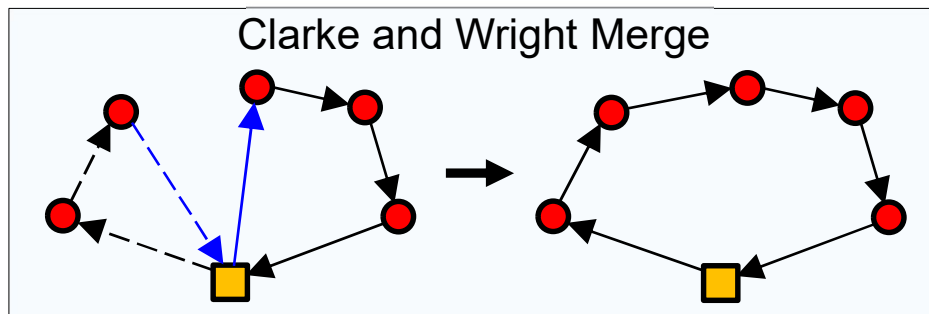
- Depot
- Customer
- Refueling station

Multi-Start Local Search Heuristic (MSLS)

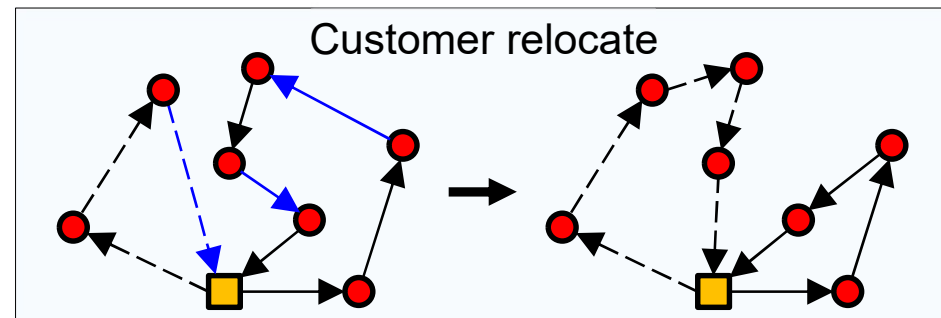
Three phases

- 1) Iteratively construct new solutions
- 2) Store vehicle routes forming these solutions in a pool \mathcal{R}
- 3) Find a set of routes in \mathcal{R} that gives least cost solution

• Example operators used in phase 1



■ Depot
● Customer



Exact algorithm

- Set partitioning formulation (SP)
 - Each possible vehicle route serves a subset of customers
 - Find least cost set of routes serving each customer exactly once

$$\begin{aligned} \text{(SP)} \quad & \min \sum_{l \in \mathcal{R}} c_l x_l \\ & \text{s.t.} \quad \sum_{l \in \mathcal{R}} a_{il} x_l = 1 \quad \forall i \in N \\ & \quad \quad x_l \in \{0,1\} \quad \forall l \in \mathcal{R} \end{aligned}$$

c_l : cost of route l

x_l : 0-1 variable equal to 1 if route l is in solution

a_{il} : 0-1 coefficient equal to 1 if route l serves customer i

\mathcal{R} : index set of all possible vehicle routes

N : set of customers

Phase 1:

- Compute lower bound **LB** by solving Linear Programming relaxation of **SP** with Subset Row [4], Weak Subset Row [1], and k -path cuts [6]
- Compute upper bound **UB** with the MSLS heuristic

Phase 2:

- Enumerate all routes \mathcal{R}^* having reduced cost $\leq \text{UB} - \text{LB}$
- Solve **SP** using only the routes in $\mathcal{R}^* \rightarrow$ **optimal solution**
- If all routes \mathcal{R}^* cannot be enumerated optimality not guaranteed

Computational results

- Benchmark problems:
 - 56 instances with 20-500 customers and 3-28 stations
- Heuristic: best new solutions to instances with 111-500 customers
 - Compared to 7 state-of-the-art heuristics [2][3][5][7][8][9]
- Exact algorithm:
 - Instances up to 111 customers 28 stations solved to optimality
 - Best exact from literature [5] solves up to 20 customer instances

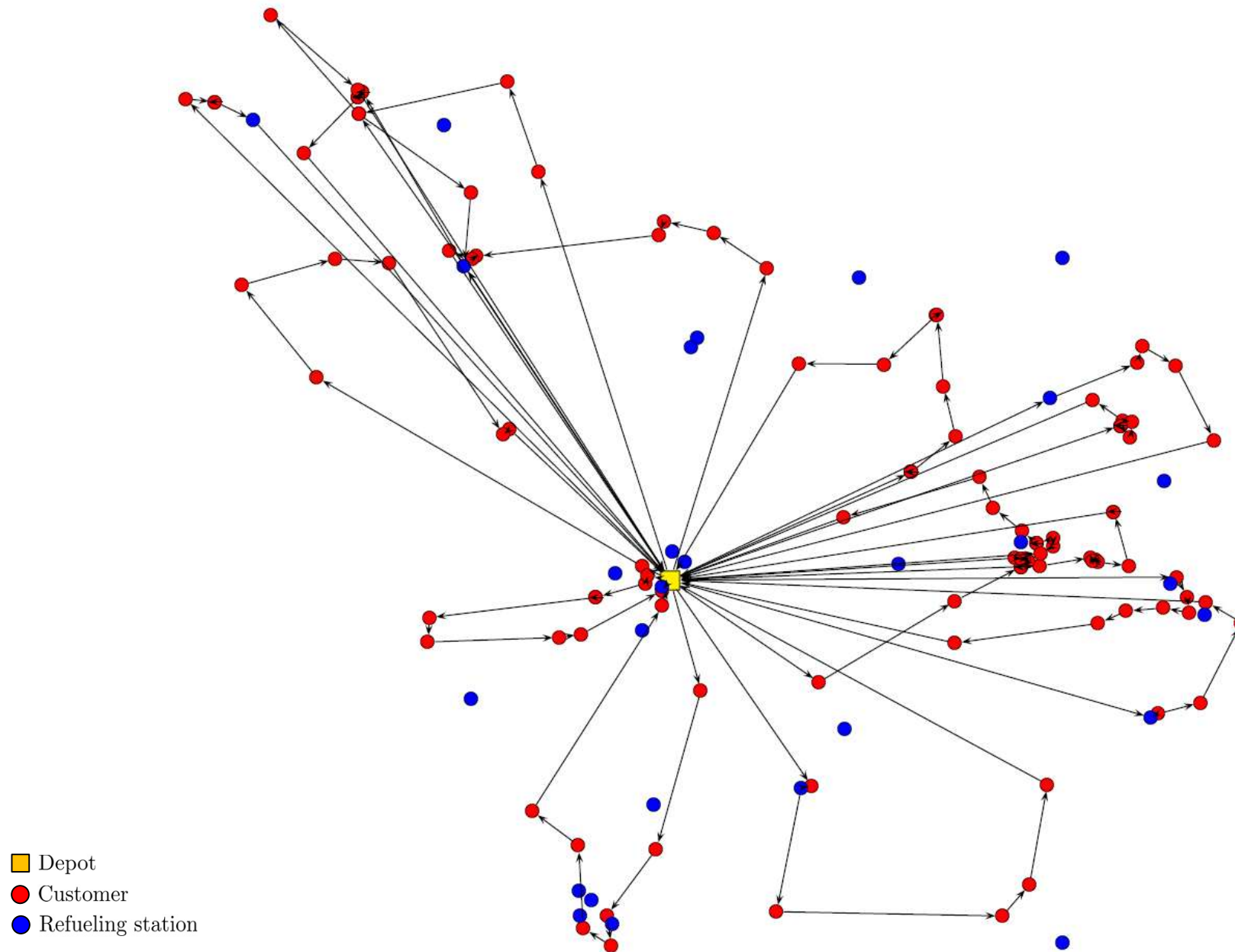
Instance name example:

75c_21s: 75 customers 21 stations

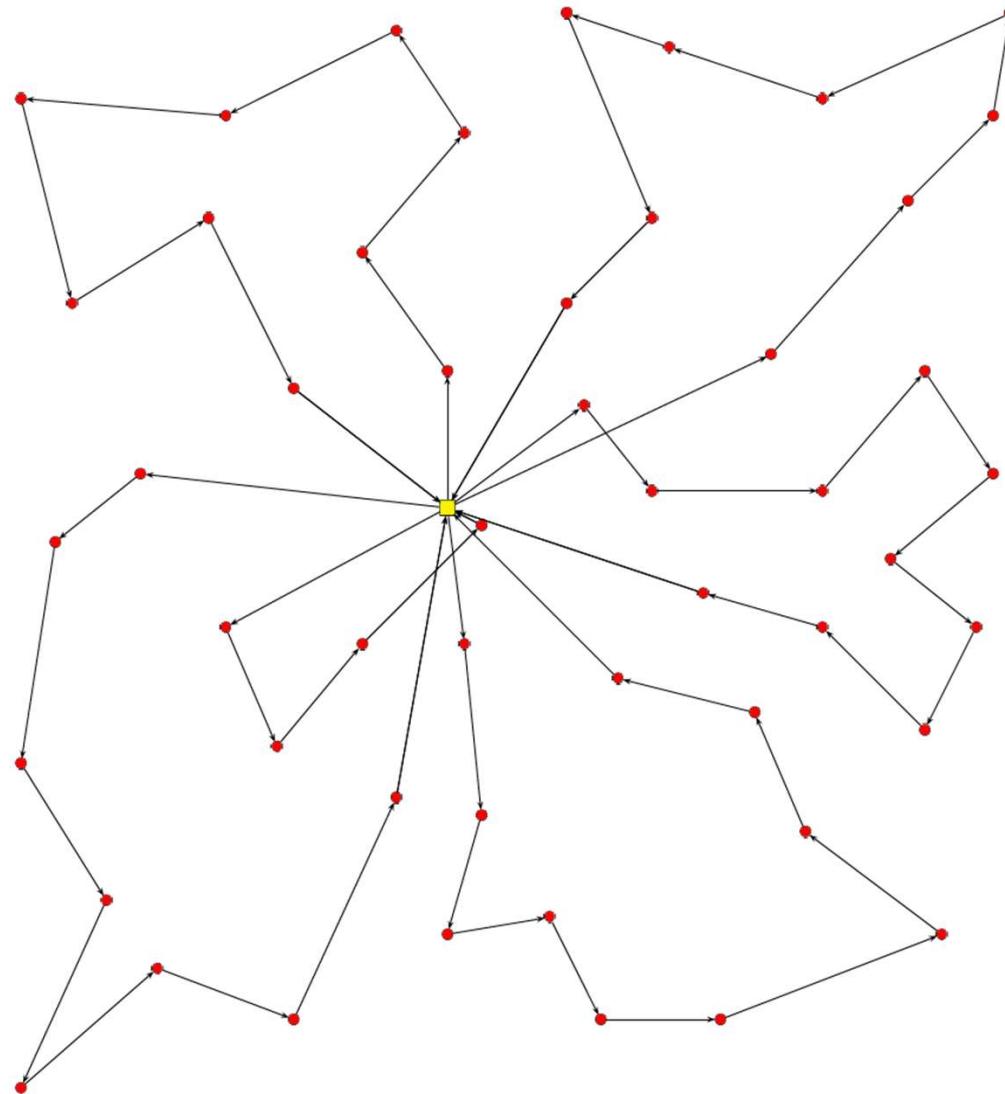
$$\%LB = \left(\frac{UB - LB}{UB} \right) * 100\%$$

Inst.	Opt	$\%LB$	Time(s)
75c_21s	*	0.00	5217
75c_28s	*	0.00	5582
100c_21s	*	0.00	5206
100c_28s	*	0.00	6531
111c_28s	*	0.13	11265
200c_21s		0.60	28135
250c_21s		0.57	27939
300c_21s		0.84	25903

Optimal solution to 111c_28s



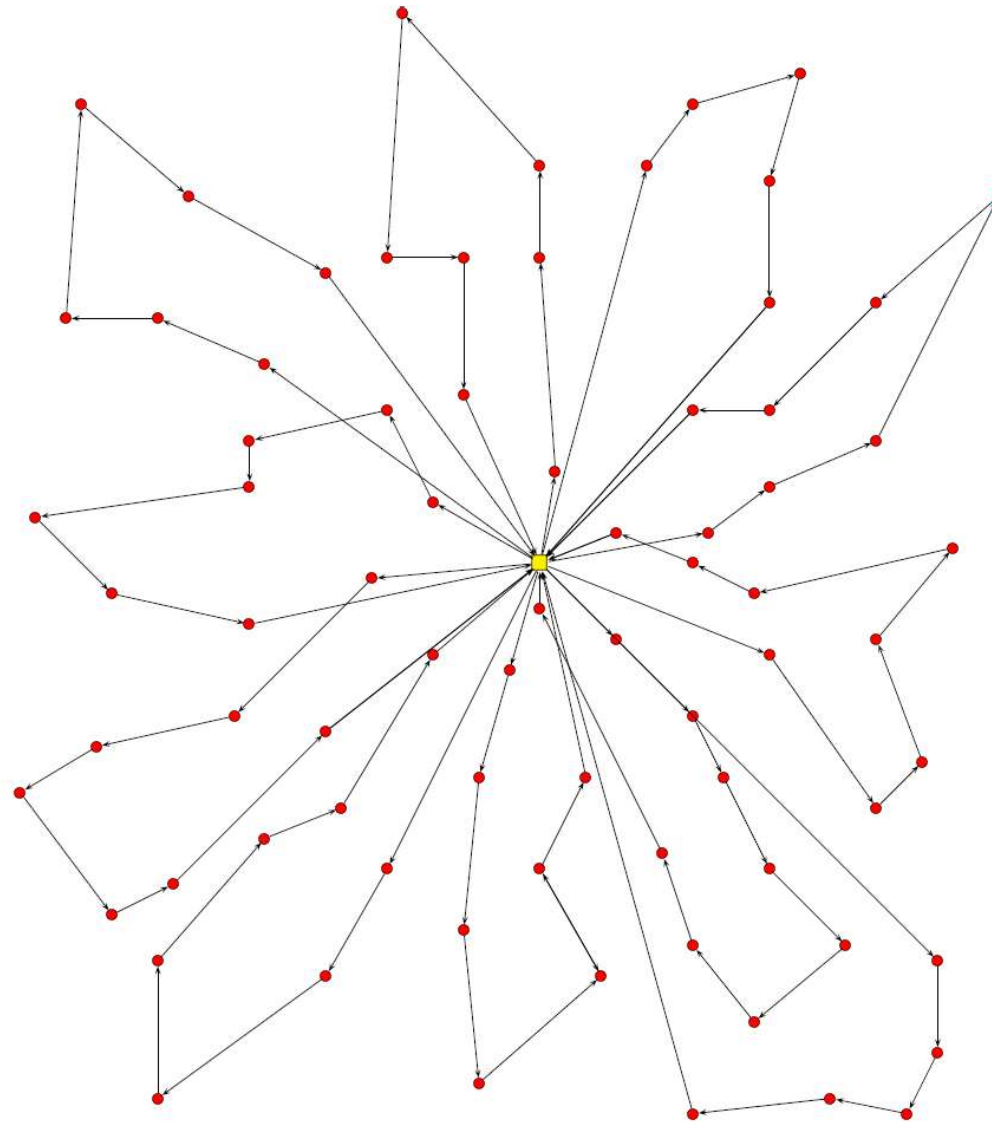
Optimal solution to Distance-constrained CVRP instance CMT6



■ Depot
● Customer

Instance	n	s	UB	Opt	$\%LB$	$\%LB_+$	Time(s)
CMT6	50	0	555.43	*	3.17	0.00	573

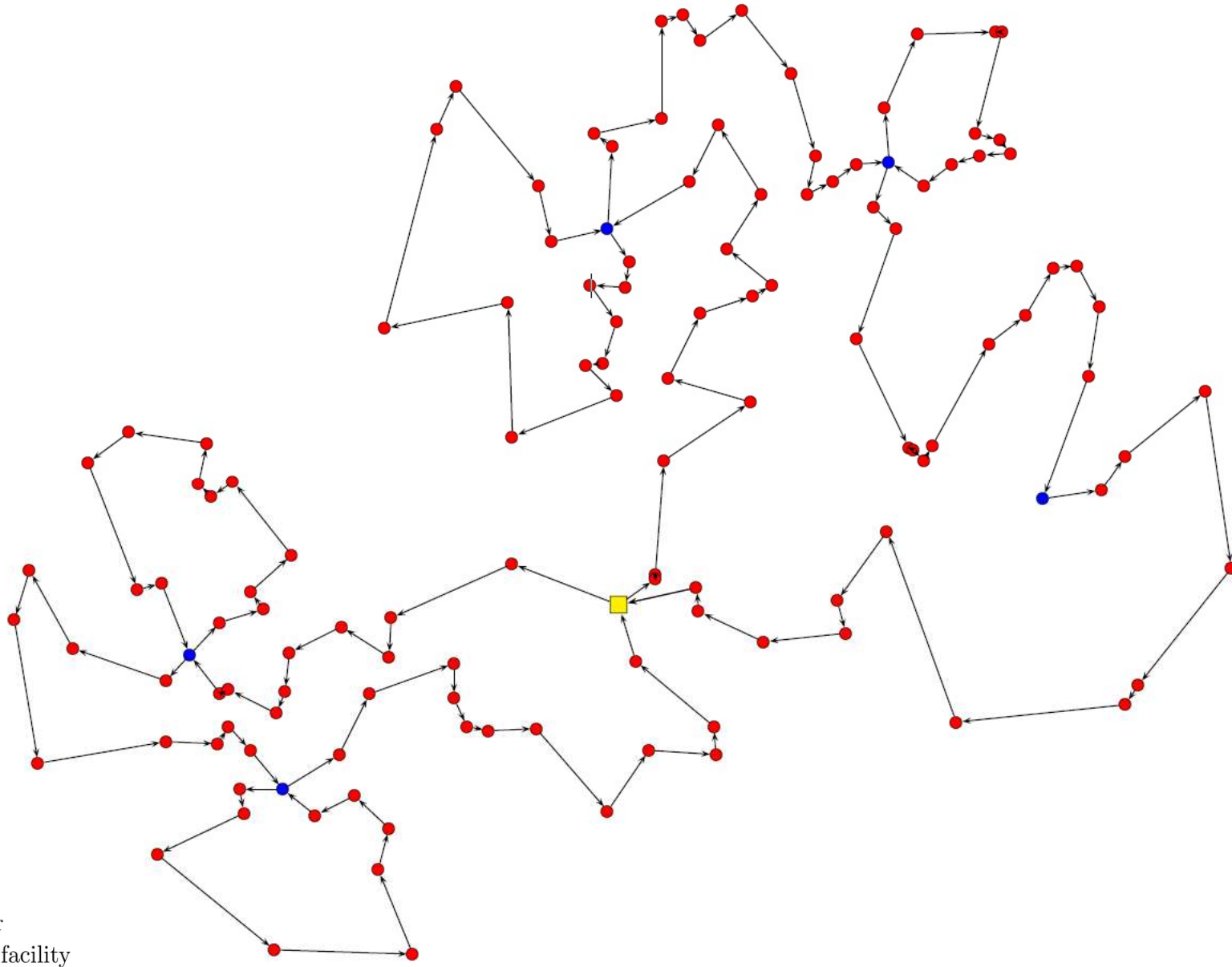
Optimal solution to Distance-constrained CVRP instance CMT7



■ Depot
● Customer

Instance	n	s	UB	Opt	$\%LB$	$\%LB_+$	Time(s)
CMT7	75	0	909.68	*	1.65	0.15	1290

Heuristic solution to VRP with satellite facilities instance



References

[1]

Baldacci, R., A. Mingozzi, R. Roberti. 2011. **New Route Relaxation and Pricing Strategies for the Vehicle Routing Problem.** *Operations Research*, 59, 1269–1283.

[2]

Erdogan, S., & Miller-Hooks, E. 2012. **A Green Vehicle Routing Problem.** *Transportation Research Part E: Logistics and Transportation Review*, 48 (1), 100–114

[3]

Felipe, A., M. T. Ortuno, G. Righini, G. Tirado. 2014. **A Heuristic Approach for the Green Vehicle Routing Problem with Multiple Technologies and Partial Recharges.** *Transportation Research Part E: Logistics and Transportation Review*, 71, 111–128

[4]

Jepsen, M., B. Petersen, S. Spoorendonk, D. Pisinger. 2008. **Subset-Row Inequalities Applied to the Vehicle-Routing Problem with Time Windows.** *Operations Research*, 56, 497–511.

[5]

Koç, Ç., & Karaoglan, I. 2016. **The green vehicle routing problem: A heuristic based exact solution approach.** *Applied Soft Computing*, 39, 154-164.

References

[6]

Laporte, G., Y. Nobert, M. Desrochers. 1985. **Optimal Routing under Capacity and Distance Restrictions.** *Operations Research*, 33, 1050–1073.

[7]

Montoya, A., C. Gueret, J. E. Mendoza, J. G. Villegas. 2015. **A Multi-Space Sampling Heuristic for the Green Vehicle Routing Problem.** *Transportation Research Part C: Emerging Technologies*, 70, 113-128

[8]

Schneider, M., A. Stenger, D. Goetze. 2014. **The Electric Vehicle Routing Problem with Time Windows and Recharging Stations.** *Transportation Science*, 48, 500–520

[9]

Schneider, M., A. Stenger, J. Hof. 2015. **An adaptive VNS algorithm for vehicle routing problems with intermediate stops.** *OR Spectrum*, 37 (2), 353-387