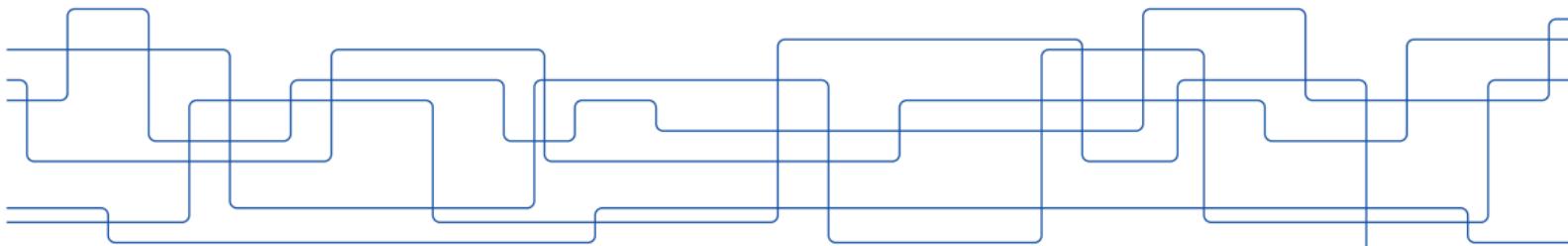




Rollout-Based Charging Strategy for Electric Trucks with Hours-of-Service Regulations

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Collaborators



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Road Freight Electrification



Charging of heavy, electric truck

Road Freight Electrification



Charging of heavy, electric truck

Positive impacts:

- 1) Reduce air and noise pollution
- 2) Mitigate climate change
- 3) Cope with energy shortages
- 4) Save operational cost
- 5) Lead to sustainable transport
- 6) ...

Truck electrification is lagging far behind...

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- ▶ Insufficient battery – Range anxiety



Limited driving range (200-600 km)

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- ▶ Drivers need to follow HoS regulations



Limited driving range (200-600 km)



	USA	EU	China
Continuous driving time (max.)	8 h	4.5 h	4 h
Mandatory rest time (min.)	30 min	45 min	20 min
Daily driving time (max.)	11 h	9 h	10 h

Hours-of-service (HoS) regulations

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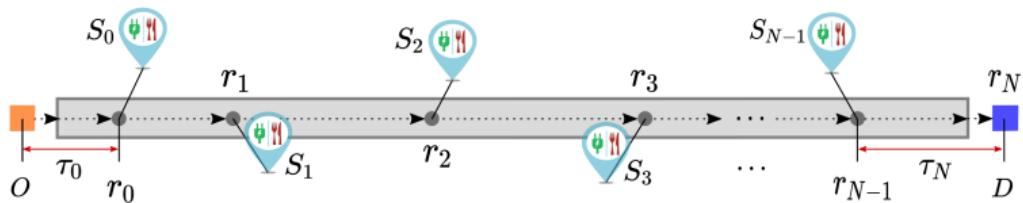


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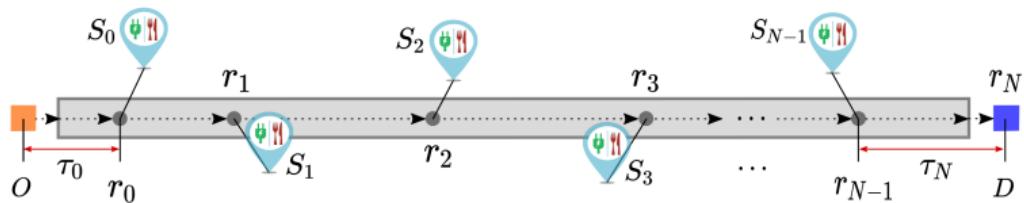
Hours-of-service (HoS) regulations

Problem: How to design reliable and efficient **charging strategies** for electric trucks to complete delivery missions on time while aligning with the HoS regulations?

Route Model



Route Model



- ▶ Decision variables:

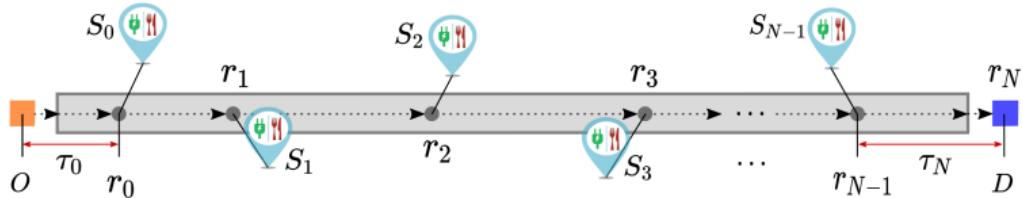
$$b_k, \tilde{b}_k \in \{0, 1\}, \quad t_k \in \Re_+$$

b_k : whether to charge at the station S_k

\tilde{b}_k : whether to rest at S_k

t_k : how long to charge the truck at S_k if $b_k=1$

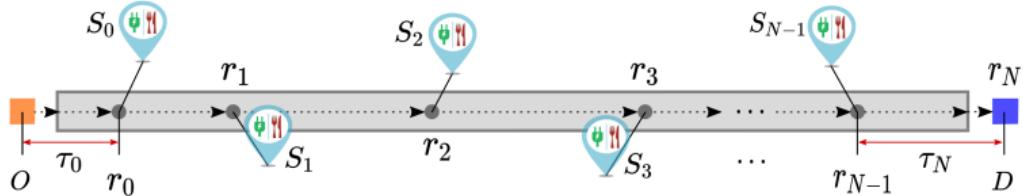
Dynamics



- ▶ The **remaining battery** upon arriving at r_{k+1} :

$$e_{k+1} = e_k + b_k \Delta e_k - \bar{P} \left(2(b_k \vee \tilde{b}_k) d_k + \tau_{k+1} \right)$$

Dynamics

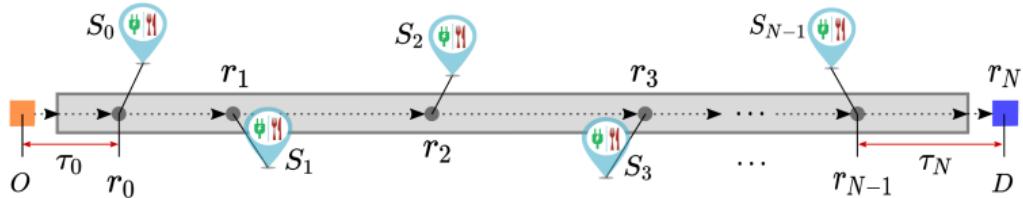


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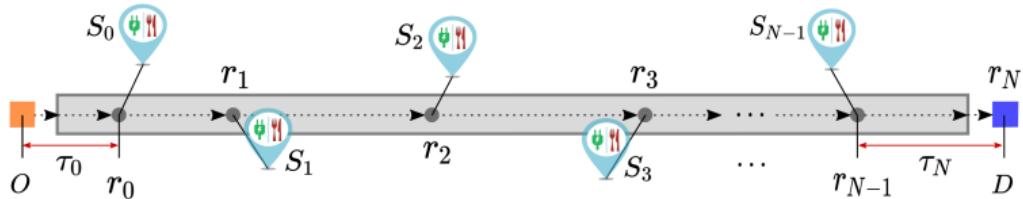
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- ▶ The **consecutive driving time** at r_{k+1} :

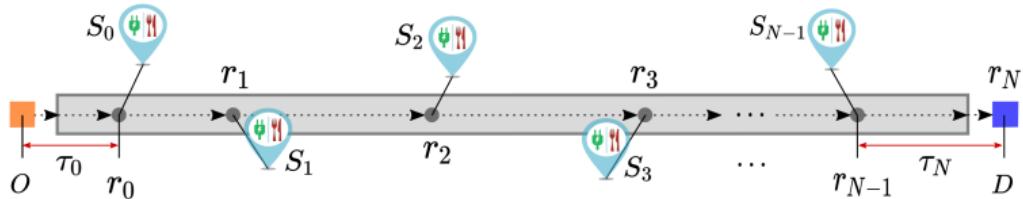
$$c_{k+1} = \tau_{k+1} + (b_k \vee \tilde{b}_k) d_k + (1 - \tilde{b}_k)(c_k + b_k d_k)$$

Constraints



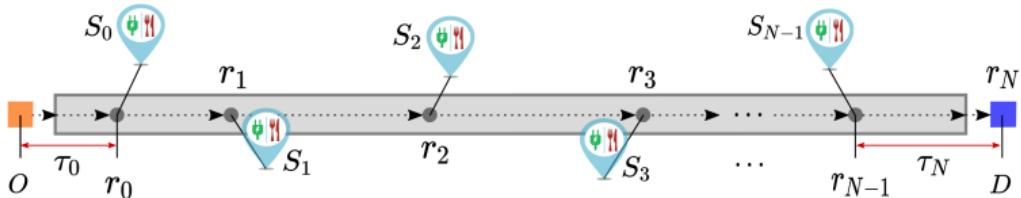
- ▶ Sufficient energy to reach S_k : $e_k \geq \text{battery for safe operation} + \bar{P}d_k$

Constraints



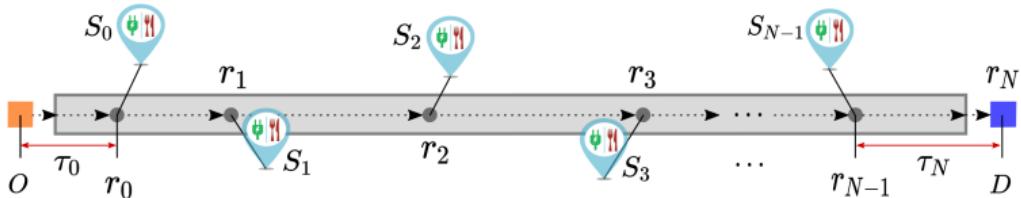
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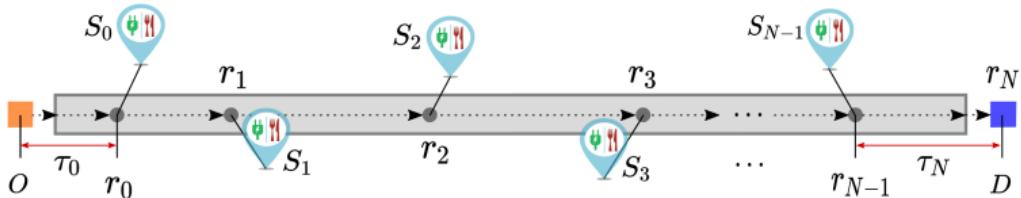
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- ▶ Delivery deadline:

$$\sum_{k=0}^{N-1} \max \left\{ \textcolor{red}{b_k}(2d_k + p_k + \textcolor{red}{t_k}), \tilde{b}_k(2d_k + T_r) \right\} \leq \Delta T$$

The Optimal Charging Problem

$$\begin{aligned} \min_{\{(b_k, \tilde{b}_k, t_k)\}_{k=0}^{N-1}} \quad & F(b_0, \tilde{b}_0, t_0, \dots, b_{N-1}, \tilde{b}_{N-1}, t_{N-1}) \\ = \sum_{k=0}^{N-1} \xi_k b_k t_k + \sum_{k=0}^{N-1} \max \left\{ b_k (2d_k + p_k + t_k), \tilde{b}_k (2d_k + T_r) \right\} \epsilon \\ \text{s. t.} \quad & \text{dynamics and constraints introduced earlier} \end{aligned}$$

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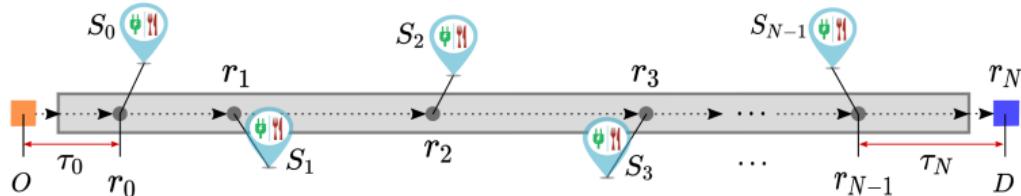
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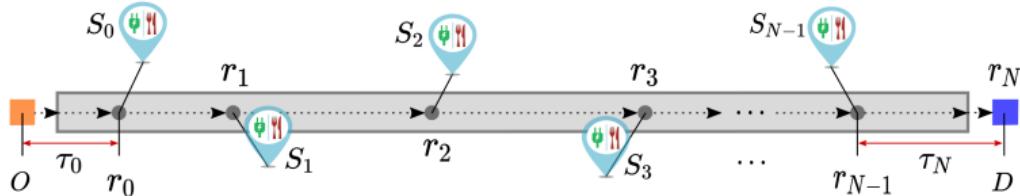
- It cannot be directly solved by many standard solvers
- Exact solutions: iterate over all possible combinations of integer variables
→ 4^N continuous optimization problems
- Linear transformation: it may still require an exponential number of iterations [see, p.480]¹

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Rollout-Based Approximate Solution



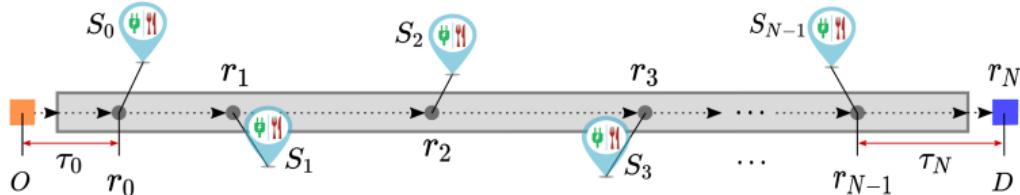
Rollout-Based Approximate Solution



► Two base solutions:

- **Greedy solution:** set $(b_k, \tilde{b}_k) = (1, 1)$ if the remaining energy is insufficient to reach S_{k+1}

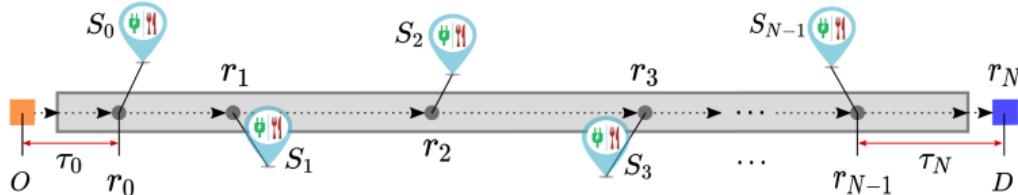
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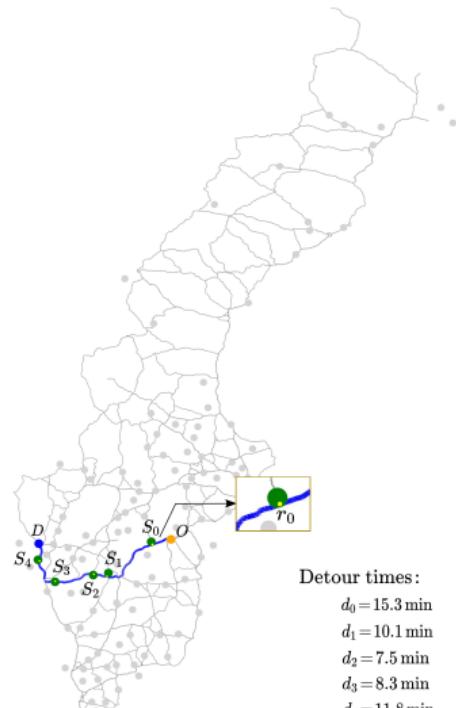
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 - **Relaxed solution:** solve a relaxation of the original problem with $b_k, \tilde{b}_k \in [0, 1]$
- ▶ Complexity: it requires solving at most $4N$ continuous optimization problems

Simulation Studies

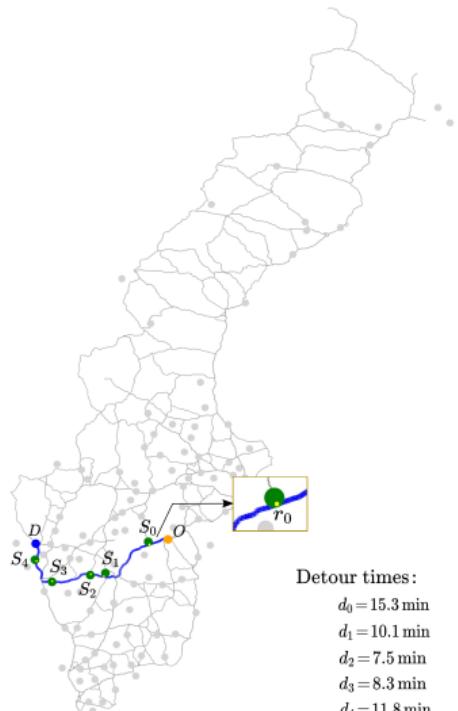


Detour times:

$d_0 = 15.3 \text{ min}$
 $d_1 = 10.1 \text{ min}$
 $d_2 = 7.5 \text{ min}$
 $d_3 = 8.3 \text{ min}$
 $d_4 = 11.8 \text{ min}$

The route of one truck.

Simulation Studies



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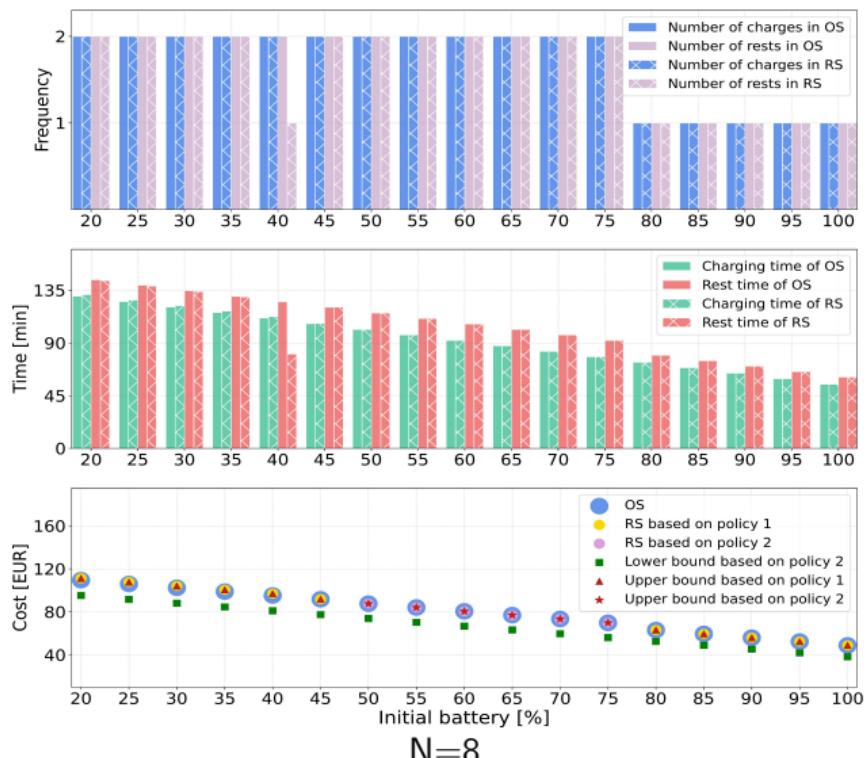
- ▶ Routes are obtained via *OpenStreetMap*
- ▶ Data for electric trucks manufactured by Scania
 - $P_k = 300 \text{ kW}$
 - $P_{\max} = 375 \text{ kW}$
 - $e_f = 468 \text{ kWh}$
 - $\bar{P} = 1.83 \text{ kWh/min}$
 - $p_k = 6 \text{ min}$
 - $\xi_k = 0.36 \text{ €/kWh}, \epsilon_k = 0.4 \text{ €/min}$
- ▶ EU's HoS regulations

Simulation Studies

- ▶ 6 scenarios (N is between 5 and 10, initial battery is between 20% and 100%)

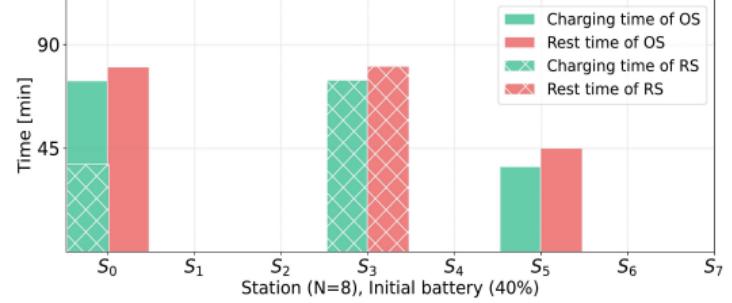
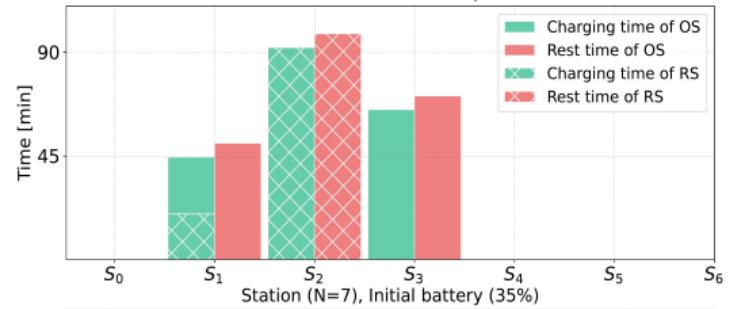
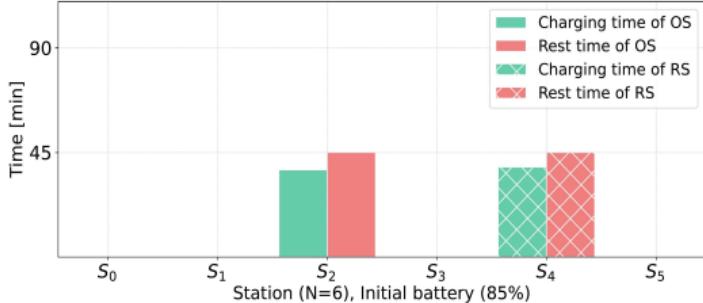
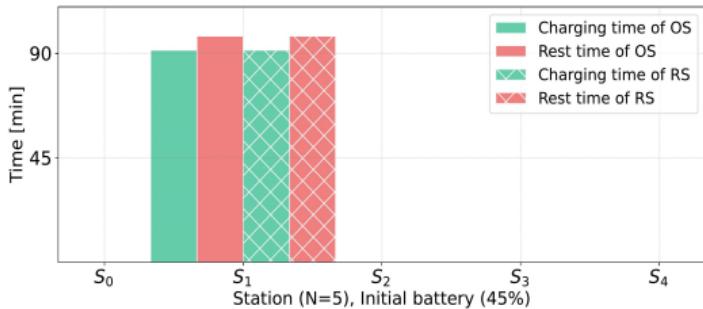
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Simulation Studies

- ▶ Comparison between the optimal solution (OS) and rollout solution (RS).



Simulation Studies

Table: Comparison between the OS and RS

N	5	6	7	8	9	10
Average optimality gap-RS [%]	0	0.55	0.72	0.49	0.03	0.42
Average optimality gap-UB [%]	1.04	5.46	2.23	0.48	4.28	1.72
Average computation time of RS [s]	0.34	0.42	0.57	0.65	0.84	1.43
Average computation time of OS [min]	0.32	1.34	5.45	24.02	98.50	413.68

- ▶ The optimality gap between the RS and OS:

$$\frac{(F(RS) - F(OS)) \times 100}{F(OS)}$$

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Future work:

- ▶ Developing optimal charging strategies with limited charging resources at stations

