

7th International Workshop on Sustainable Road Freight Transport

Modeling Strategic and Operational Policy Decisions for EV Sharing Platforms

Vishal Bansal^a, Deepak Prakash^b, K.B. Devika^b, Debjit Roy^a, Shankar C. Subramanian^b

a: Indian Institute of Management Ahmedabad, India

b: Indian Institute of Technology Madras, India

Motivation

Benefits of Electric Vehicles (EVs):

- Around 24% of CO₂ emissions are contributed to by the transportation section due to IC engine vehicles.
- EVs offer zero tailpipe emissions, better efficiency over IC engine vehicles, and reduces reliance on fuel.
- 17 countries have announced 100% zero-emission vehicle through 2050.

Adoption Issues and Motivation:

- In spite of advantages, penetration of EVs is very less (around 1% of global fleet of cars is electrified).
- Issues for adoption include
 - Range anxiety
 - Cost of EVs
 - Inadequate charging infrastructure
- Use of EV sharing platforms can help alleviate these issues.
- In India, in 2025, 17% of cars are expected to be sold to fleet owners, and the number of shared rides to increase by three times (from 2018).

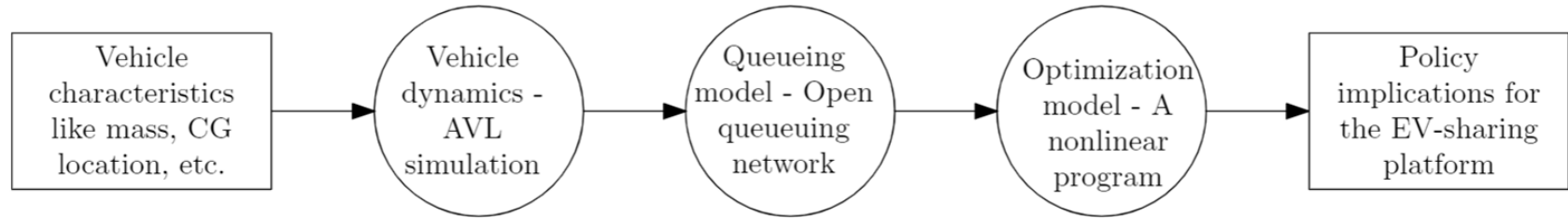
Literature Review

| Research stream | Study | EV-sharing platform | Station-based system | Charging levels | Traffic conditions | Vehicle dynamics | Powertrain and regenerative braking model |
|---|--|---------------------|----------------------|-----------------|--------------------|------------------|---|
| Performance analysis of vehicle-sharing platforms | George and Xia (2011), Roy et al. (2014) | | Y | | | | |
| | Chen et al. (2016), He et al. (2017), Guo et al. (2018), Loeb et al. (2018), Hua et al. (2019), He et al. (2020) | Y | | | | | |
| EV modeling | Shao et al. (2018) | Y | | | | Y | |
| | Dandl and Bogenberger (2019) | Y | | Y | Y | | |
| | Alesiani and Maslekar (2014) | Y | | | | Y | |
| | Chen et al. (2018) | Y | | Y | | Y | |
| Our work | | Y | Y | Y | Y | Y | Y |

Research Questions

1. How to integrate the vehicle and network dynamics with the optimization of design parameters of EV-sharing platform?
2. How to analyze the effect of different traffic conditions on the decrease in the battery energy of the vehicle while traveling?
3. How does the consideration of powertrain and regenerative braking models impact the estimates of decrease in the battery energy of the vehicle?
4. How do partial charging and vehicle exit from the platform influence the platform's profitability?

Overall Analysis Framework



- Stage 1 - Vehicle dynamics model to calculate the energy drawn from the battery per unit distance for different traffic conditions
- Stage 2 - Open queueing network to model the EV-sharing platform and its operations
- Stage 3 - Optimization model to determine the optimal system parameters

EV Modeling

Tractive force from the powertrain (electric motor) needs to overcome 4 resistive forces:

1. Rolling resistance
2. Aerodynamic drag
3. Grade resistance
4. Inertia

Single motor drive configuration was considered in this study.

Motor modeling:

1. Ideal motor characteristics used to represent continuous torque-speed profile.
2. Efficiency map used to include motor and inverter losses.

Regenerative braking modeling:

1. Series regenerative braking for optimum braking performance chosen.
2. Braking strategy ensures ideal brake force distribution is followed, hence ensures safety.

For front-wheel drive configuration, distribution between friction brakes and motor is given by

$$F_{bf,fric}(t) = \begin{cases} 0, & F_{bf,ideal}(t) < F_{regen,max}(t) \\ F_{bf,ideal}(t) - F_{regen,max}(t), & F_{bf,ideal}(t) \geq F_{regen,max}(t) \end{cases},$$
$$F_{br,fric}(t) = F_{br,ideal}(t),$$

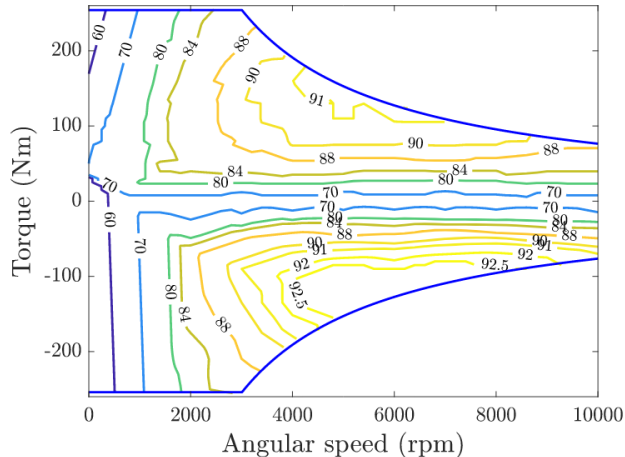
where

$$F_{regen,max}(t) = \begin{cases} -\frac{\tau_{max}(\omega(t))GR}{r\eta_t}, & \omega(t) \geq \omega_{base,cont} \\ 0, & \omega(t) < \omega_{base,cont} \end{cases}.$$

Case Study - Nissan e-NV200 Evalia

Specifications of vehicle considered:

- Light vehicle with single motor front-wheel drive configuration.
- Battery modelled using an Open Circuit Voltage vs State of charge curve and an internal resistance.
- Constant accessory power consumption of 1.4 kW considered.



| Specification | Meaning | Value |
|-----------------------------|---|---------|
| M_{laden} (kg) | Mass of the fully laden vehicle | 2250 |
| $M_{unladen}$ (kg) | Mass of the fully unladen vehicle | 1592 |
| L (mm) | Wheelbase | 2725 |
| $l_{f,laden}$ (mm) | Distance of CG from the front axle center for the fully laden vehicle | 1429.12 |
| v_{max} (km/h) | Maximum longitudinal vehicle speed | 123 |
| r (mm) | Tyre rolling radius | 310.75 |
| E_{rating} (kWh) | Energy rating of the fully charged battery | 40 |
| n_{cells} | Total number of cells | 192 |
| Q_{cell} (Ah) | Rated cell capacity | 56.3 |
| R_{int} (Ω) | Internal cell resistance | 0.002 |
| P_{cont} (kW) | Continuous power rating of the motor | 80 |
| τ_{cont} (Nm) | Continuous torque rating of the motor | 254 |
| ω_{max} (rpm) | Maximum angular speed of the motor | 10000 |
| $\omega_{base,cont}$ (rpm) | Continuous base angular speed of the motor | 3008 |
| GR | Gear ratio of the final drive | 9.301 |
| h_{laden} (mm) | Height of CG from the ground for the fully laden vehicle | 800 |
| C_d | Coefficient of drag | 0.35 |
| A_f (m ²) | Frontal projected area | 2.8043 |
| ρ (kg/m ³) | Mass density of air | 1.225 |
| η_t | Transmission efficiency | 0.95 |
| f_r | Coefficient of rolling resistance | 0.01 |

EV Model Validation and AVL CRUISE Results

Model Validation:

- Vehicle simulated using the World Harmonized Light Vehicle Test Procedure.
- Energy consumption obtained with EV model (= 0.2633 kWh/km) matched closely the quoted value by Nissan (= 0.2591 kWh/km).

Traffic conditions considered:

| Traffic condition | Drive cycle | Parameters | | | |
|-------------------|-------------|------------------|-----------------|------------------------|------------------------|
| | | s_{cycle} (km) | t_{cycle} (s) | $v_{max,cycle}$ (km/h) | $v_{avg,cycle}$ (km/h) |
| Low | HWFET | 16.45 | 765 | 96.40 | 77.57 |
| Medium | UDDS | 12.07 | 1369 | 91.20 | 31.50 |
| High | ECE-15 | 1.013 | 195 | 50 | 19 |

Results from AVL CRUISE:

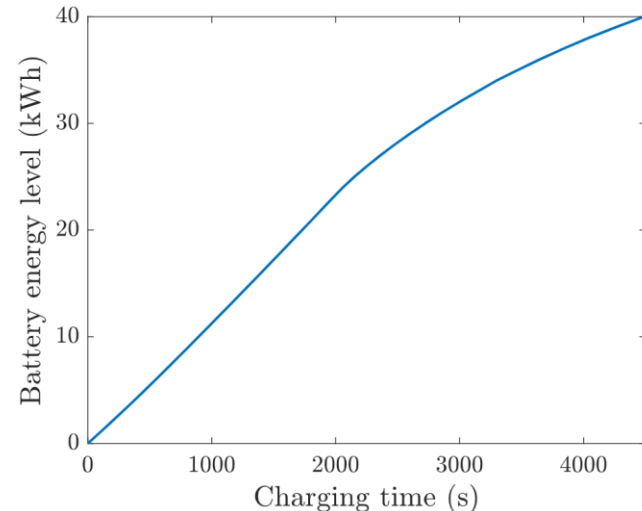
$\Delta E_{battery}$ over HWFET, UDDS, and ECE-15 are 0.2375, 0.2725, and 0.2771 kWh/km.

Comparison with simplified EV model:

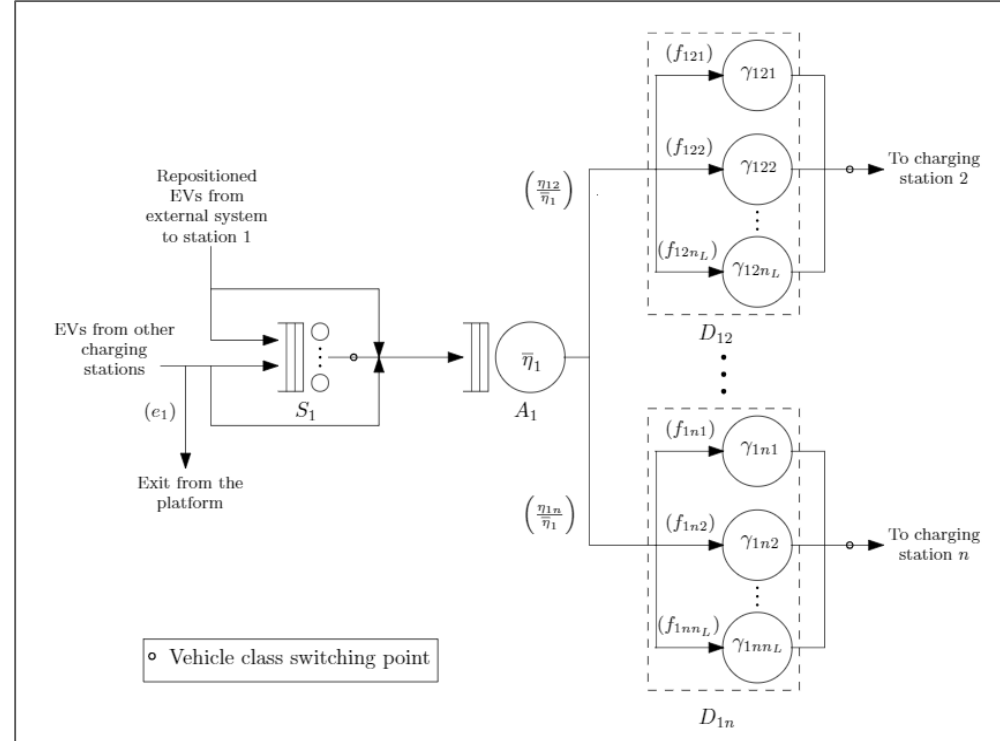
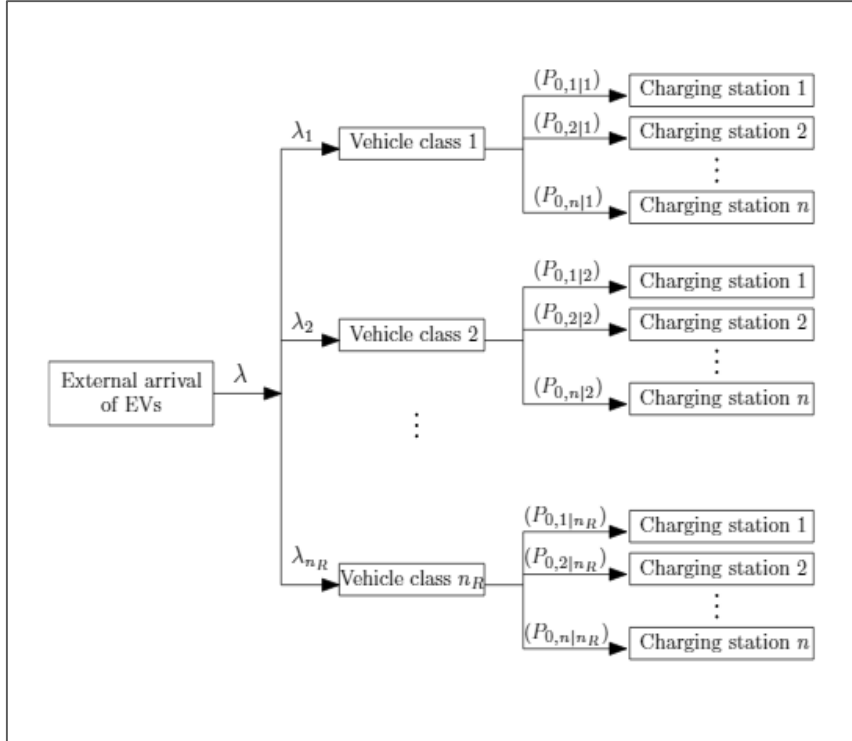
- Net efficiency factor of 80%.
- 20% regenerative braking factor.

$\Delta E_{battery}$ over HWFET, UDDS, and ECE-15 are 0.2233, 0.2605, and 0.2593 kWh/km.

Battery energy vs charging time:



Queuing Network of the EV-sharing Platform



Optimization Model for Setting Policy Parameters

- Mixed-integer nonlinear optimization problem (MINLP)
- Decision variables:
 - Number of chargers at each charging station
 - Distribution of external arrival of EVs to different charging stations
 - EV queue length for charging
 - EV queue length for trip assignment
 - Queuing node utilization
 - EV flow rates between different queuing nodes

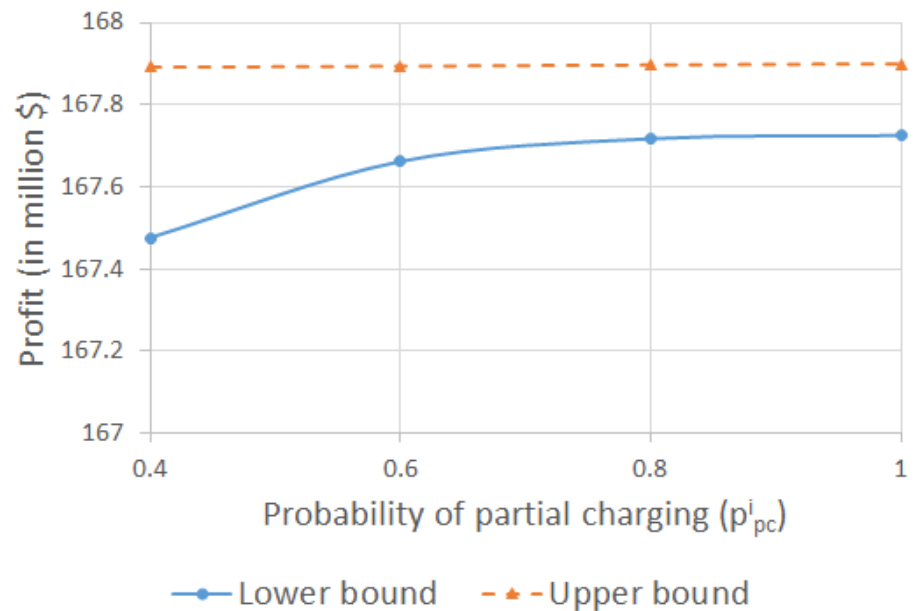
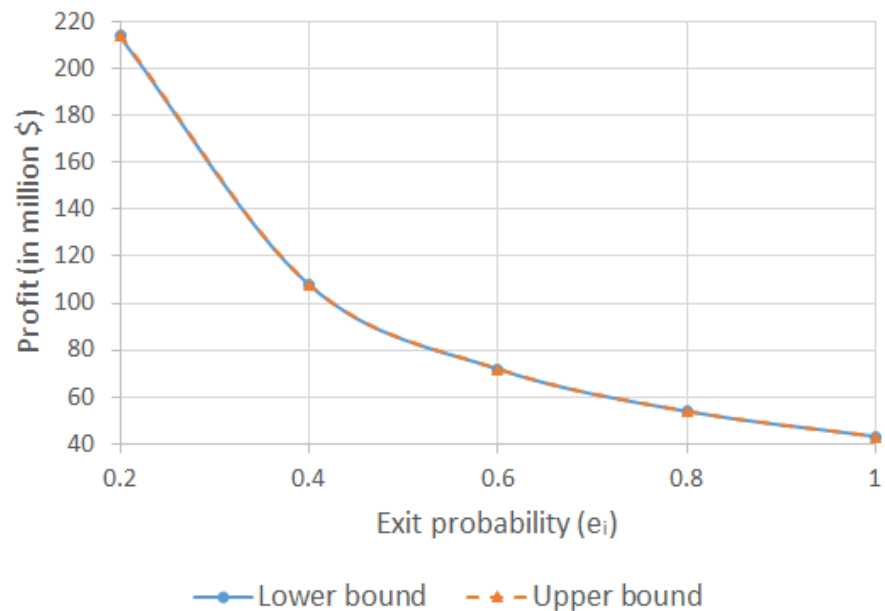
Optimization Model for Setting Policy Parameters

- Objective function: Maximize the platform's annual profit
 - Revenue from customer trips
 - Charger installation cost
 - Waiting cost of the EVs at charging stations
 - Repositioning cost of the EVs to charging stations
- Constraints:
 - Sum of fractions of repositioned EVs of a particular vehicle class to different stations is 1
 - Flow-balance constraints at the queuing nodes
 - Queuing node utilization constraints
 - Queue length constraints
 - Capacity constraints for the charging station

Solution Method and Results

- MINLP is nonlinear and non-convex due to integer variable and fractional constraints with multi-linear cubic and quadratic terms.
- Constraint for the decision variable - EV queue length for charging is causing multi-linear cubic terms.
- Bound-based heuristic
 - Overestimate the decision variable to obtain the lower bound of the optimal profit
 - Underestimate the decision variable to obtain the upper bound of the optimal profit
- The optimality gap between the lower and upper bounds from the heuristic, is found to be less than 0.5%.

Managerial Insights



Contributions and Future Directions

- We propose an integrated analytical framework to address the operational and infrastructural challenges faced by an EV-sharing platform.
- We provide a bound-based heuristic to solve a mixed-integer nonlinear optimization model with fractional constraints and multi-linear cubic terms.
- Our analysis provides various operational insights for the policy makers of the EV-sharing platform.
- Possible extensions:
 - Heterogeneous EV fleet with different vehicle loading conditions
 - A more specific and realistic vehicle speed profile by collecting real time data
 - Joint determination of optimal partial charging probabilities and target energy level for partial charging along with the number of chargers

References

- D. K. George and C. H. Xia, "Fleet-sizing and service availability for a vehicle rental system via closed queueing networks," *European journal of operational research*, vol. 211, no. 1, pp.198–207, 2011.
- D. Roy, J. A. Pazour, and R. De Koster, "A novel approach for designing rental vehicle repositioning strategies," *IIE Transactions* , vol. 46, no. 9, pp. 948–967, 2014.
- F. Guo, J. Yang, and J. Lu, "The battery charging station location problem: Impact of users' range anxiety and distance convenience," *Transportation Research Part E: Logistics and Transportation Review* , vol. 114, pp. 1–18, 2018.
- L. Chen, L. He, and Y. H. Zhou, "Managing electric vehicle charging: An exponential cone programming approach," Available at SSRN , 2020.
- T. D. Chen, K. M. Kockelman, and J. P. Hanna, "Operations of a shared, autonomous, electric vehicle fleet: Implications of vehicle & charging infrastructure decisions," *Transportation Research Part A: Policy and Practice* , vol. 94, pp. 243–254, 2016.
- B. Loeb, K. M. Kockelman, and J. Liu, "Shared autonomous electric vehicle (saev) operations across the Austin, Texas network with charging infrastructure decisions," *Transportation Research Part C: Emerging Technologies* , vol. 89, pp. 222–233, 2018.
- L. He, H.-Y. Mak, Y. Rong, and Z.-J. M. Shen, "Service region design for urban electric vehicle sharing systems," *Manufacturing & Service Operations Management* , vol. 19, no. 2, pp. 309–327, 2017.
- Y. Hua, D. Zhao, X. Wang, and X. Li, "Joint infrastructure planning and fleet management for one-way electric car sharing under time-varying uncertain demand," *Transportation Research Part B: Methodological* , vol. 128, pp. 185–206, 2019.
- L. He, G. Ma, W. Qi, and X. Wang, "Charging an electric vehicle-sharing fleet," *Manufacturing & Service Operations Management* , 2020.
- S. Shao, W. Guan, and J. Bi, "Electric vehicle-routing problem with charging demands and energy consumption," *IET Intelligent Transport Systems*, vol. 12, pp. 202–212, 2018.
- F. Dandl and K. Bogenberger, "Comparing future autonomous electric taxis with an existing free-floating carsharing system," *IEEE Transactions on Intelligent Transportation Systems* vol. 20, no. 6, pp. 2037–2047, 2019.
- F. Alesiani and N. Maslekar, "Optimization of charging stops for fleet of electric vehicles: A genetic approach," *IEEE Intelligent Transportation Systems Magazine*, vol. 6, pp. 10–21, 2014.
- T. Chen, B. Zhang, H. Pourbabak, A. Kavousi-Fard, and W. Su, "Optimal routing and charging of an electric vehicle fleet for high-efficiency dynamic transit systems," *IEEE Transactions on Smart Grid*, vol. 9, no. 4, pp. 3563–3572, 2018.

Thank You!

Questions/Comments?