SlrpEV Smart Learning Research Pilot for Electric Vehicles

Project Overview Apr 4, 2023







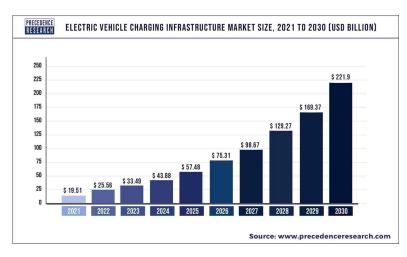


EV Charging Technology & Market





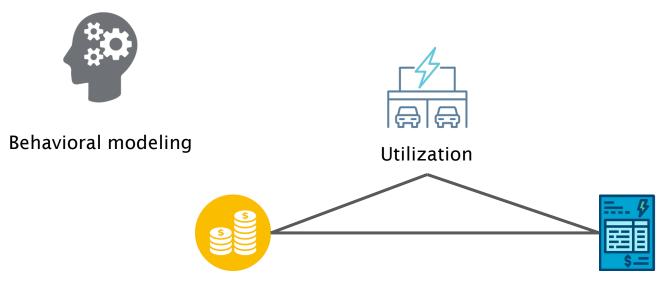
• 50,000 public EV charging stations in the U.S., with a total of nearly 130,000 individual (EVSE) charging ports



 EV Charging Infrastructure Market Projected Growth to \$ 221.9 billion by 2030 from \$25.56 billion in 2022 at CAGR of 31.02%



Revenue Maximization at Workplace Charging Facilities





Menu of options: Regular and Scheduled charging



Trade-off between:

Gross Revenue

- 1. Price of charging and charger utilization
- 2. Decreasing cost and charger utilization









Literature Review & Existing Projects

- ☐ Smart-Charging Projects
 - ☐ OptimizEV Project in Upstate New York
 - ☐ ChargeTO in Toronto

Residential place only Fixed pricing options

- ☐ Literature
 - ☐ Control the deadline of users(A. Ghosh, 2019);
 - □ Nudge users to connect to more profitable chargers(N. Tucker, 2018).

Historical dataset only Cannot reflect user behaviour and sensitivity





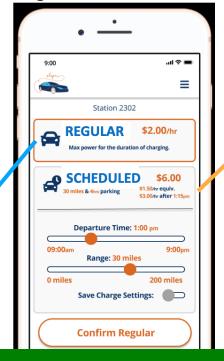
Cyber-Physical & Human System

→ 3 sites: UC Berkeley, UC San Diego, SunPower San Jose; 26 EVSEs fully operational since

Nov 2020

REGULAR

- Fixed rate in USD/hr
- Max power until unplug or top-off
- Charging power is uncontrollable.



SCHEDULED

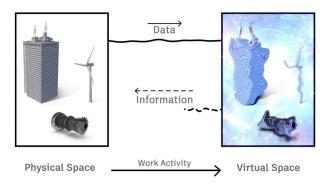
- User provides departure time & added range
- Total cost fixed *a* priori
- Charging power is controllable.

Charging schedules & pricing dynamically optimized based on congestion, electric demand, energy prices, user choices, etc.

SlrpEV Simulation

- Digital Twin
 - A digital twin is a virtual model of a physical object.
 - uses real-time data sent to simulate the behavior and monitor operations
 - **□** Objective
 - Data Ingestion
 - Evaluation of the charging station model
 - Data Analysis

THE DIGITAL TWIN MODEL



SlrpEV Simulation flow chart

Session Generation

Goal

Retrieve random arrival, departure, and energy-needed data from the actual facility data. $(11/2020 \sim 1/2022)$

Input

Actual charging station facility data

Output

- · 10 users per day
- 10 days Station-of level user data
- · Arrival time
- · Departure time
- · Energy needed

Data Pre-processing

Goal

Receive Session Generator's input data frame and filter out unnecessary data

Input

Sessions generated from the session generator

Output

Filtered Session data

- Overnight charging scenario
- Energy needed overpasses the maximum charging limit.
- Pre-process arrival and departure data to conform with minimum time interval (15mins)

Main Simulation

Goal

Overall simulation. run the simulation until the given runtime (10 days)

Input

Integrate with station-level optimizer

· Session data (arrival time, departure time, energy needed)

Output

Filtered Session data

- Charger assignment if all the chargers are assigned, "unavailability leave"
- · Choice of each user
- · User choice and (Regular /Scheduled) charging price
- Daily list of user's (Regular /Scheduled / Leave) choices.

Station-level Optimizer

Input

· Session data

Output

- · User choice probability
- (Regular / Scheduled) charging price

Simulator Choice Function

Input

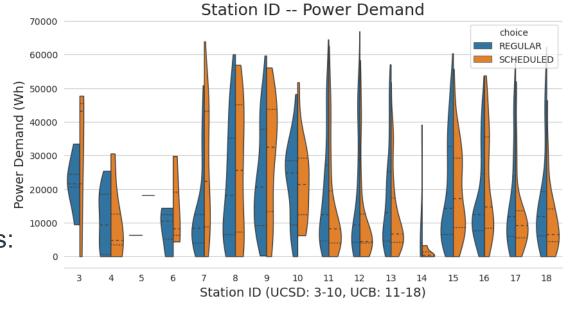
· Charging price

Output

• User's charging mode choice (Regular / Scheduled / Leave)



- User Trends:
 - o Time
 - Demand
 - Choice
 - 0
- Supplier Trends:
 - Different Costs
 - 0

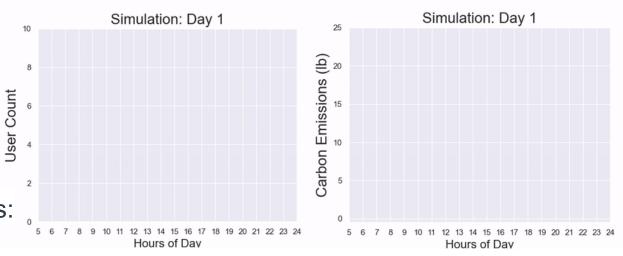




- User Trends:
 - Numbers
 - Emissions
 - Choice
 - 0

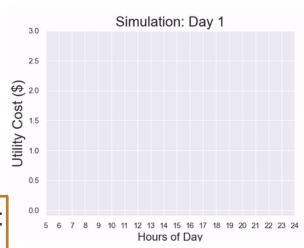


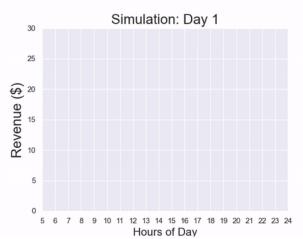
- Different Costs
- 0





- User Trends:
 - Numbers
 - Emissions
 - Choice
 - 0
- Supplier Trends:
 - Different Costs
 - 0

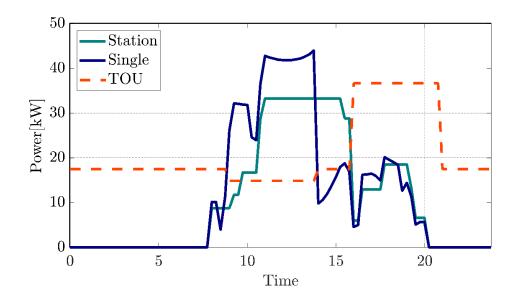






• Station optimization:

- Station
- Demand
- Choice
- 0



	Single-charger	Station-level
Max Power(kW)	43.98	33.23



Conclusion

Overview

- Simulator/Optimizer
- Data Analysis

Next Steps

- Deployment
- More chargers

Opportunities

- o Relationship between supply and demand
- Industry growth
- 0



References

- [1] P. Alexeenko and E. Bitar, "Achieving reliable coordination of residential plug-in electric vehicle charging: A pilot study," 12 2021. [Online]. Available: http://arxiv.org/abs/2112.04559.
- [2] J. Bauman, M. Stevens, S. Hacikyan, L. Tremblay, E. Mallia, and C. Mendes, "Residential smart-charging pilot program in toronto: Results of a utility controlled charging pilot," World Electric Vehicle Journal, vol. 8, pp. 531-542, 6 2016.
- [3] A. Ghosh and V. Aggarwal, "Control of charging of electric vehicles through menu-based pricing," IEEE Transactions on Smart Grid, vol. 9, pp. 5918-5929, 11 2018.
- [4] N. Tucker and M. Alizadeh, "Online pricing mechanisms for electric vehicle management at workplace charging facilities." IEEE, 10 2018, pp. 351-358.

