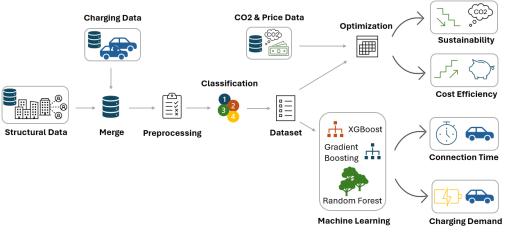
## **Master's Thesis of Sebastian Stein**

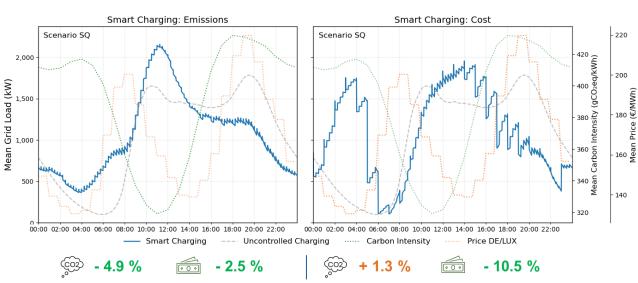
## **Mentoring:**

M.Sc. Mario Ilic M.Sc. Markus Fischer



Graphical representation of modeling framework

The findings indicate a moderate potential for improved sustainability and saving costs, while SC generally positively affects the distribution of the energy load. CO2optimized charging generally depletes the typical two peaks and instead produces a larger peak of 2.15 MW that is shifted one hour from the morning peak to 11:30 h. As a result, the charging demand around the evening peak, where prices are highest during the day, is also reduced. In contrast, cost-driven charging shifts a large share of charging to night times between 01:00 to 05:00 h, coinciding with the lowest energy prices.



charging behavior using machine learning models.

The rapid adoption of electric vehicles (EVs) is a pivotal strategy

for mitigating climate change and reducing greenhouse gas

emissions in the transport sector. However, the growing electric vehicle (EV) fleet poses significant challenges for energy grids,

necessitating innovative solutions to manage increased electricity

demand effectively. This thesis focuses on the potential of smart

charging (SC) strategies to address these challenges within the

context of public EV charging infrastructure in Munich. Using

detailed charging session records provided by Stadtwerke

München (SWM) for 2021 to 2022, along with structural data, historical CO2 intensity of the German electricity mix, and spot

market energy prices, we analyze the status quo of public EV

charging behavior in Munich. Key objectives include quantifying

ecological and techno-economic parameters, assessing the

potential of two SC strategies, and evaluating the predictability of

Potential of Smart Charging at public EV charging stations in Munich

IM\_HOUR\_OF\_DAY **Connection Time** T\_AVG\_KWH\_AT\_HOUR cat VEHICLE TYPE EV 37 VEHICLE\_TYPE\_EV\_11 MAE 1.58 h \_CS\_AVG\_KWH\_AT\_HOUP AVG DUR H PER CS SMAPE 49.5 % num\_DAY\_OF\_WEEK cat VEHICLE TYPE EV 46 R2 0.45 num\_CS\_ID 0.25 0.05 0.10 0.15 0.20 Importance 0.30 m\_VT\_AVG\_KWH\_AT\_HOUR VEHICLE TYPE EV 37 um\_VT\_AVG\_DUR\_AT\_HOUR im\_CS\_AVG\_KWH\_AT\_HOUR **Charging Demand** cat VEHICLE TYPE EV 11 MAE 4.42 kWh \_\_\_AVG\_KWH\_PER\_C AVG\_DUR\_H\_PER\_CS **SMAPE** 44.6 % num CS ID num\_\_DAY\_OF\_WEEK at\_\_VEHICLE\_TYPE\_EV\_74 R2 0.29 um\_HOUR\_OF\_DAY VEHICLE TYPE EV 46 0.15 nportan 0.05 0.20 0.25 0.10

\_CS\_AVG\_DUR\_AT\_HOUR

Results and feature importance of machine learning prediction

This thesis proposes a machine learning framework to identify and understand patterns in the charging behavior of EVs, which is essential for improving scheduling strategies. The best predictive models demonstrate that temporal and vehicle-specific patterns are central in predicting charging behaviors. Our predictive accuracy falls short of those in the literature from which we derive suggestions for improvement and propose to use our results as a benchmark for future development. We therefore consider this thesis as a contribution to the broader understanding of the importance of localized, data-driven approaches and advanced modeling techniques needed to fully realize the potential of public EV charging infrastructure in Munich, aligning ecological goals with techno-economic feasibility.

