

Charging ahead

Firm-level optimisation strategies for sustainable and cost-effective electric vehicle workplace charging

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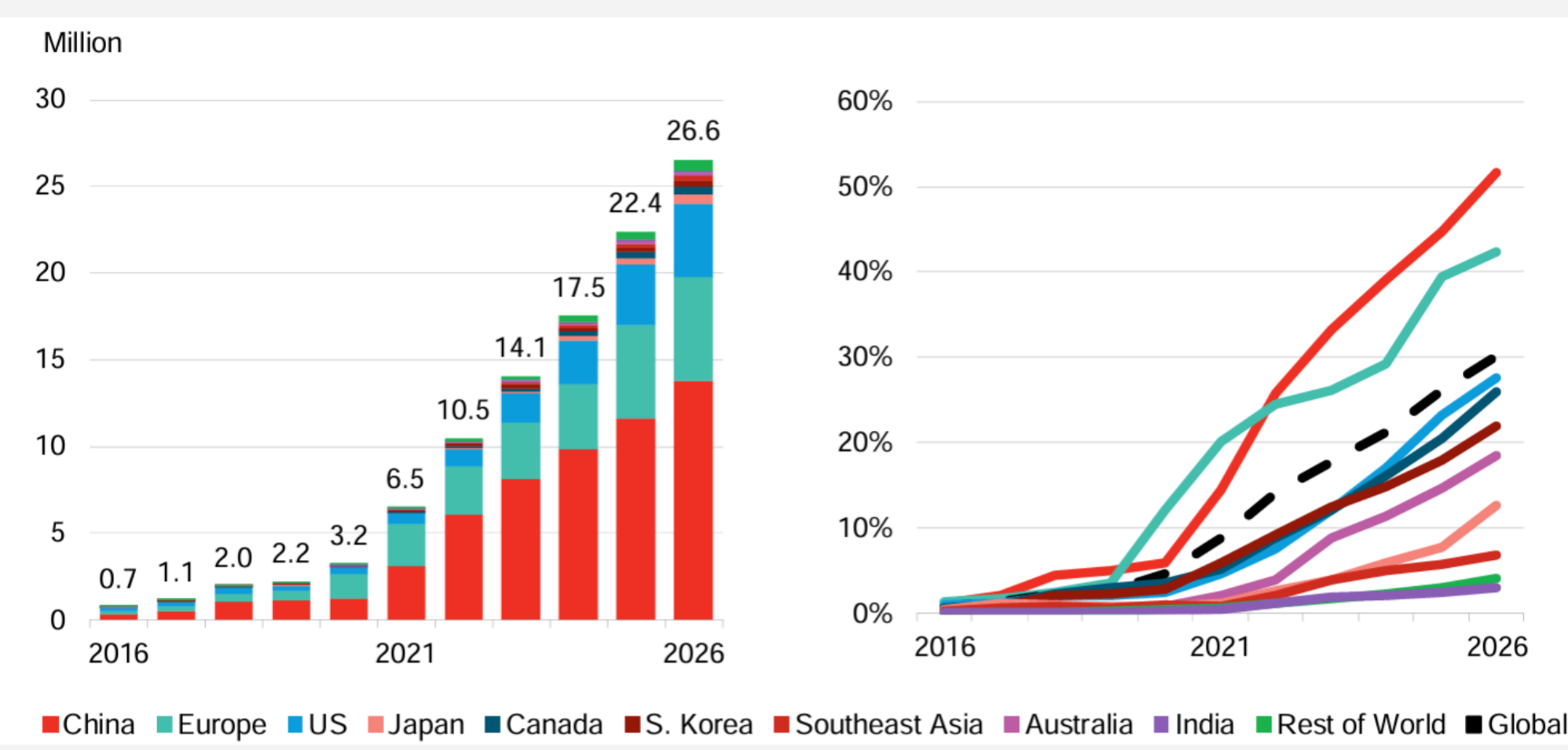


SoGE School of Geography and the Environment



1 | Background & Motivation

Problem context: Decarbonising transport



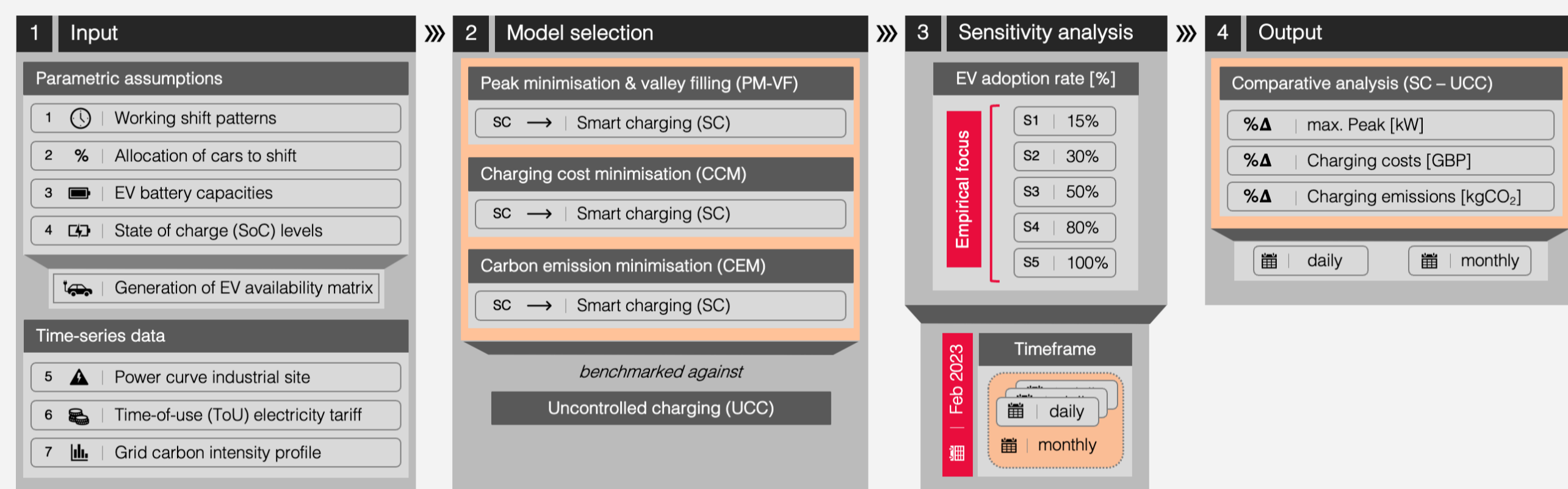
Global near-term EV sales (l.) and share of new passenger vehicle sales by market (r.) [1].

- 88% of GHG emissions are covered by net-zero legislation as of 2023 [2].
- Mitigation efforts in transport sector feature strong focus on road vehicle electrification.
- 65% of commitments in nations' revised nationally determined contributions (NDCs) as of the Glasgow Climate Pact (2021) are focused on electrification & fuel-switching [3].
- Helping deliver these commitments requires widespread charging infrastructure at workplaces and public places to bring 'convenience parity' between EVs and internal combustion vehicles (ICVs) [4].

2 | Model Structure

Approach: Outlining four-step structure

Topic | Leveraging digital technologies to unlock demand-side flexibility (DSF) of electric vehicles (EVs)



RQ | What are the benefits of coordinated EV workplace charging for firms?

3 | Methodology

Methods: Drawing from operations research (OR) [5, 6]

Peak min. & valley filling (PM-VF):

$$\min z_{PM-VF} = \sum_{t \in T} (P_t + y_t - C)^2$$

Charging cost min. (CCM):

$$\min z_{CCM} = \sum_{t \in T} y_t * \lambda_t$$

Carbon emission min. (CEM):

$$\min z_{CEM} = \sum_{t \in T} y_t * \gamma_t$$

[1] s.t. $y_t = \sum_{m \in M} x_{mt} f_{mt} \quad \forall t \in T$ Total charging load

[2] $-p_{max} \leq x_{mt} \leq p_{max} \quad \forall t \in T; m \in M$ Charging power restrictions

[3] $0 \leq E_m^{ini} + \sum_{k \in T | k \leq t} x_{mk} f_{mk} \leq E_m^{cap} \quad \forall t \in T; m \in M$ Battery capacity restrictions

[4] $E_m^{fin} = E_m^{ini} + \sum_{k \in T | k \leq t} x_{mk} f_{mk} \geq E_{T+1} \quad \forall t \in T; m \in M$ Minimum state-of-charge (SoC) requirement

[5] $0 = x_{mt}(1 - f_{mt}) \quad \forall t \in T; m \in M$ Logical operator ensuring car availability

$C = \frac{\max(P_t) + \min(P_t)}{2}$ Constant C

$f_{mt} = \begin{cases} 1, & \text{if EV } m \in M \text{ is parked at workplace at time } t \in T, \\ 0, & \text{otherwise} \end{cases}$ Definition of car availability matrix

Modelling assumptions

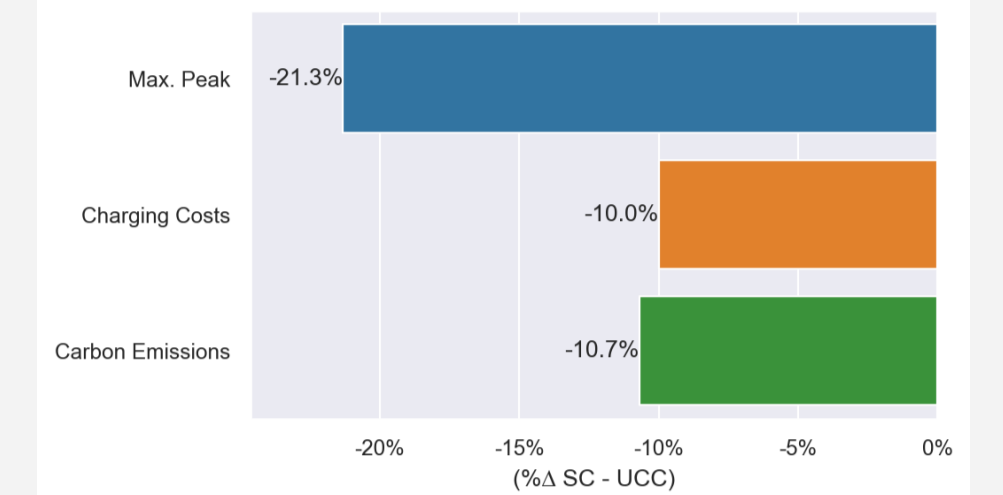
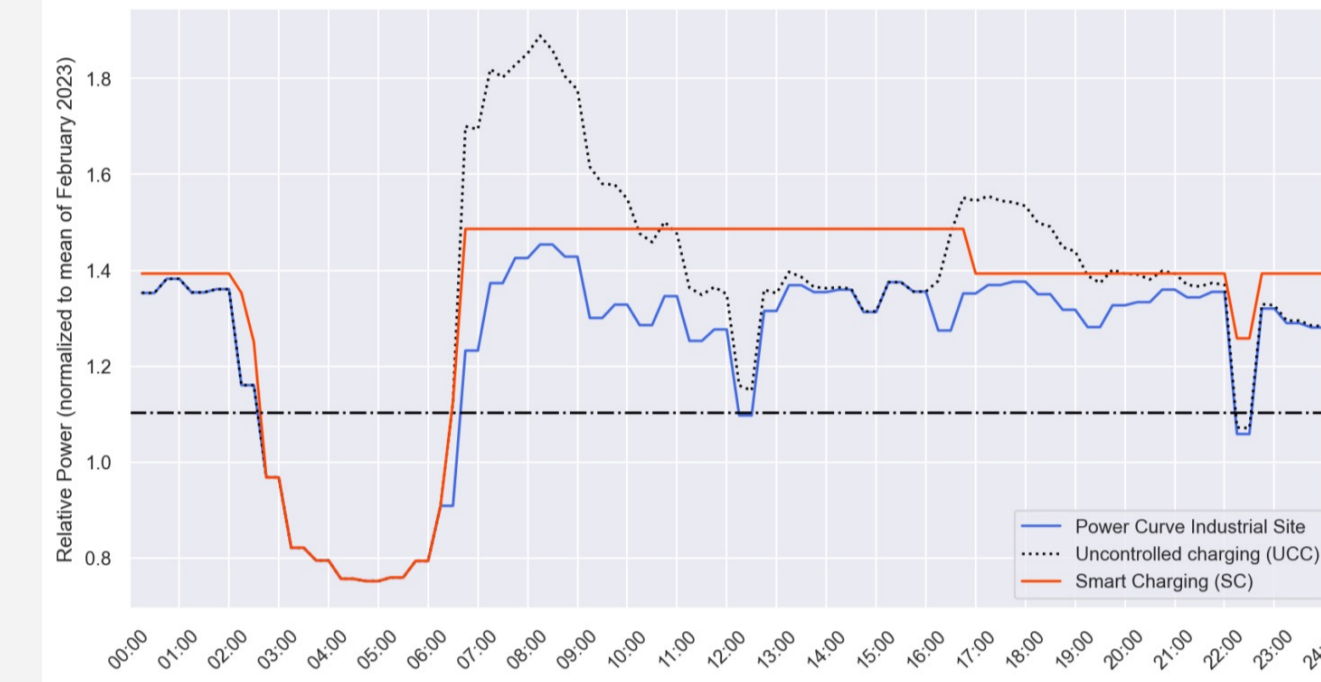
Relative allocation of cars				EV battery capacity			State of Charge (SoC) levels		
AM	PM	OFFICE	#CARS	kWh	kWh	kWh	LB	UB	
63%	27%	10%	1,100	48	71	100	E_{ini}	10% 80%	
							E_{fin}	80% 100%	

uniformly distributed

4 | Results

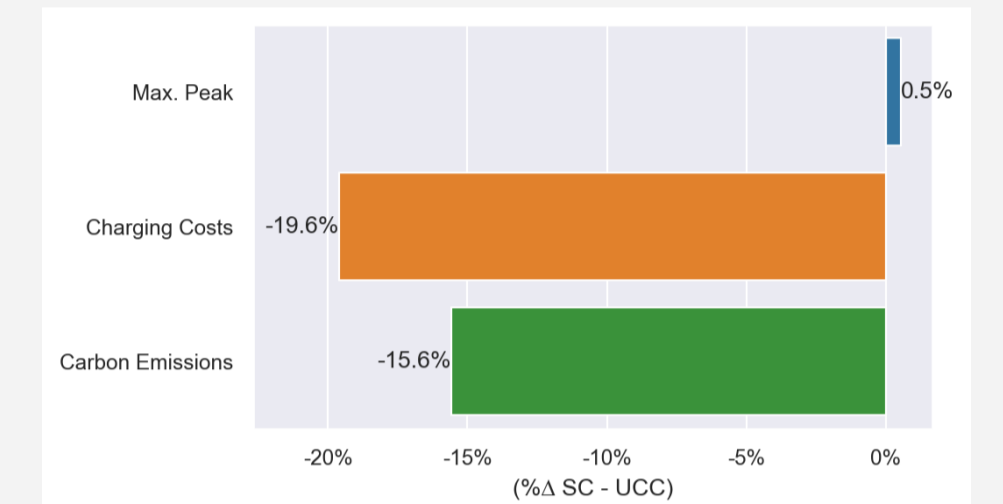
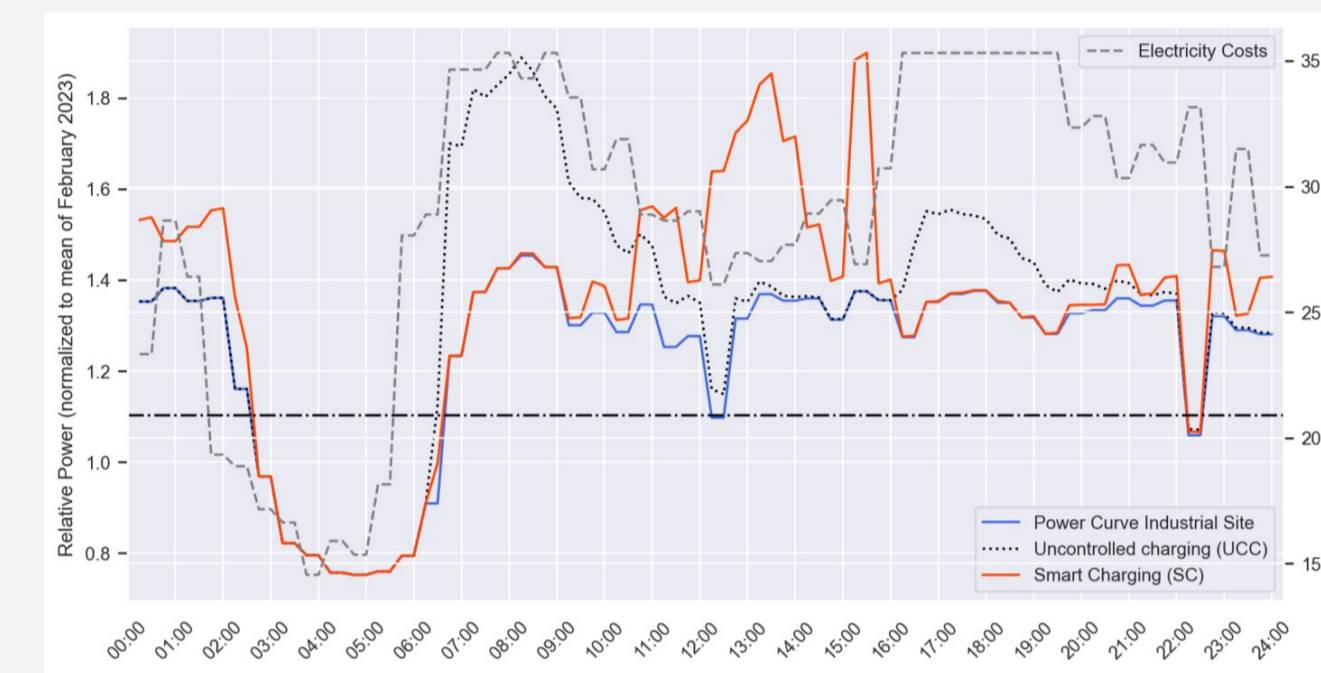
Analysis: Sensitivity analyses (exemplarily for EV adoption rate = 50%)

Peak minimisation & valley filling (PM-VF):



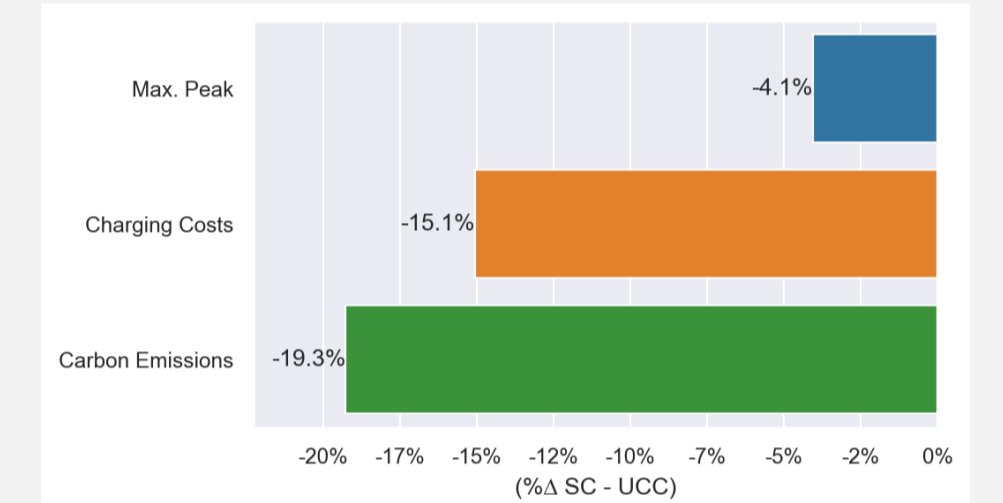
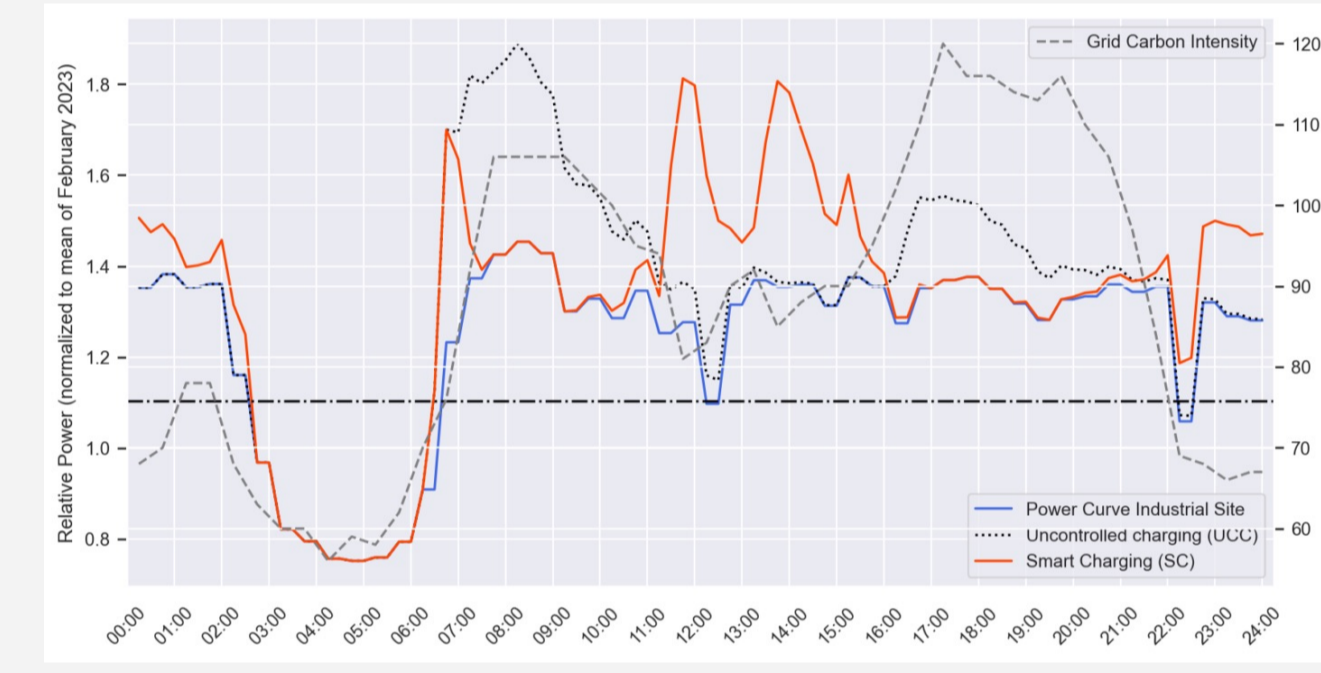
Relative performance of PM-VF (change in output (SC - UCC [%Δ]))*

Charging cost minimisation (CCM):



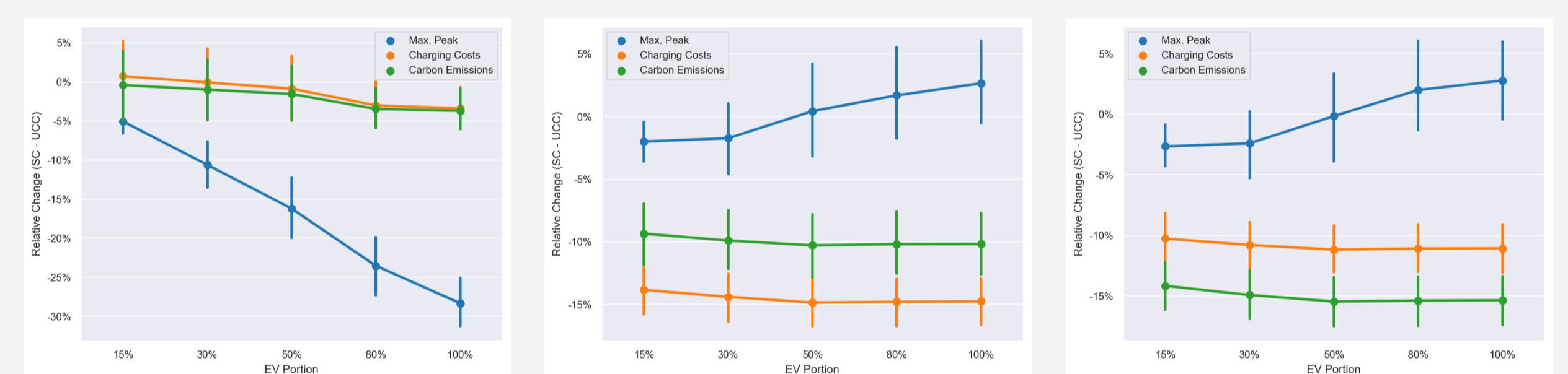
Relative performance of CCM (change in output (SC - UCC [%Δ]))*

Carbon emission minimisation (CEM):



Relative performance of CEM (change in output (SC - UCC [%Δ]))*

* smart charging (SC) / uncontrolled charging (UCC)



Visual summary of key metrics max. peak, charging costs and carbon emissions differentiated by model type | EV rates [S1-5: 15-100%]. Quantitative assessment of output changes [in %Δ], measured against UCC, for PM-VF (l.), CCM (m.), and CEM (r.), exemplarily for 01 February, 2023.

5 | Web Application

Development of interactive open-source tool

EV Workplace Charging Dashboard

Input Parameters: Shift Patterns, EV Battery Capacities, Grid Carbon Intensity Profile.

Model selection: Peak Minimisation & Valley Filling (PM-VF), Charging Cost Minimisation (CCM), Carbon Emission Minimisation (CEM).

Results: Relative performance of PM-VF, CCM, and CEM (change in output (SC - UCC [%Δ])).

Data pipeline: python, PYOMO, GUROBI OPTIMIZATION, Streamlit.

Model formulation: PYOMO.

Optimisation: GUROBI OPTIMIZATION.

Visualisation: Streamlit.

GitHub: anticipated release Q4/2024, github.com/segermarcel.

try it yourself: QR code.

[1] BloombergNEF (2023). Electric Vehicle Outlook 2023.

[2] UN Environment Programme (2023). Emissions Gap Report.

[3] SLOCAT (2021). Climate Strategies for Transport. An analysis of NDCs and Long-Term Strategies.

[4] Dixon et al. (2020). On the ease of being green. Applied Energy 258.

[5] Ioakimidis et al. (2018). Peak shaving and valley filling. Energy 148.

[6] Zheng et al. (2019). Integrating plug-in EVs into power grids. Renewable and Sustainable Energy Reviews 112.

To cite

Seiger et al. (2024). Charging ahead: Firm-level optimisation strategies for sustainable and cost-effective electric vehicle workplace charging. ICT4S Conference. Stockholm, Sweden.

Case study involving



Acknowledgments

