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The Impact of the Weather on Electric Vehicle Fleet Demand

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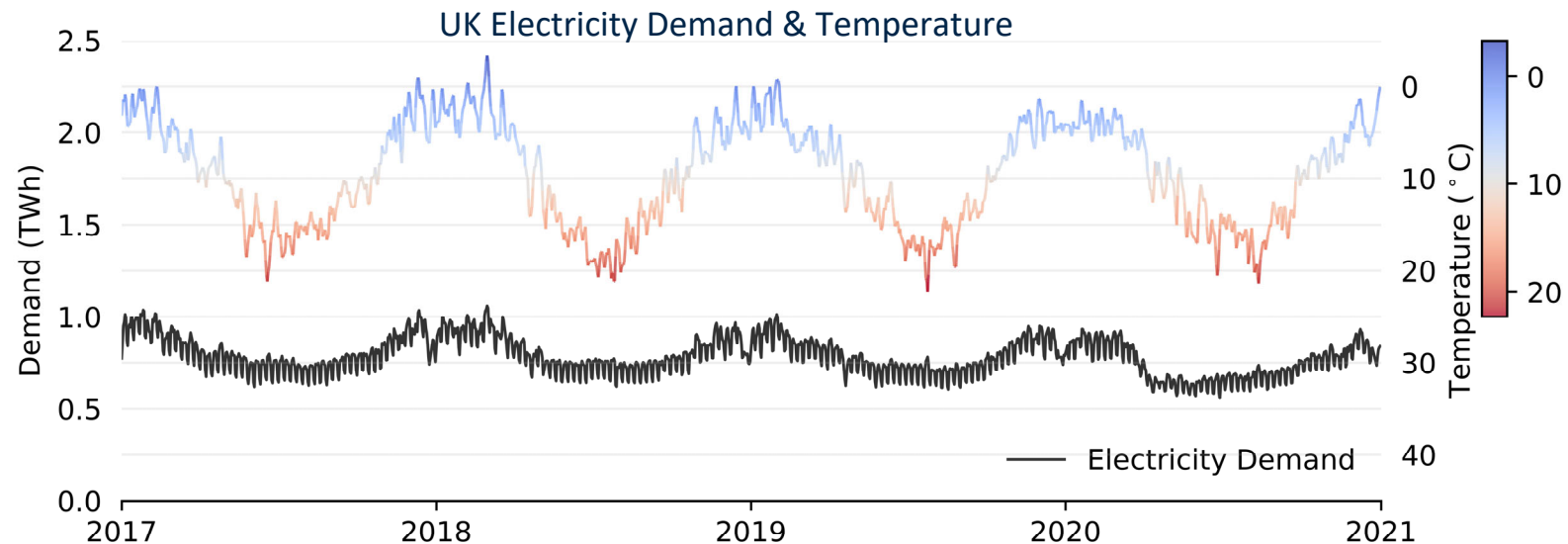
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1. Evidence for how EV energy consumption is impacted by the weather
 2. Two methods to model this relationship
 3. Case study: German fleet
 4. Case study: UK fleet
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Weather has an increasing impact on electricity supply and demand

➤ *Staffell & Pfenninger (2018)*

- Electricity supply and demand are becoming **increasingly weather-dependent**
- Heat electrification could see **peak demand increase by 20% in 15 years**
- Additional weather driven loads may worsen this issue, **increasing cost or risk security of supply**



There is only limited data of how weather impacts EV demand

- Anecdotal evidence
 - NAF (2020) *Twenty popular EVs tested in Norwegian winter conditions*
 - EVs on average lose **20 percent** of their range in colder climate
 - EVs **charge more slowly** in cold temperatures
 - 1. **Yuksel & Michalek (2015)** *Effects of Regional **Temperature** on Electric Vehicle Efficiency, Range, and Emissions in the United States; and 7 additional sources.*
 - 2. **National Grid & Element Energy (2019)** *EV charging data; 8 million UK charging events in 2017.*
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EV demand can be modelled as a function of the weather

1. Fleet demand as a function of **ambient temperature**

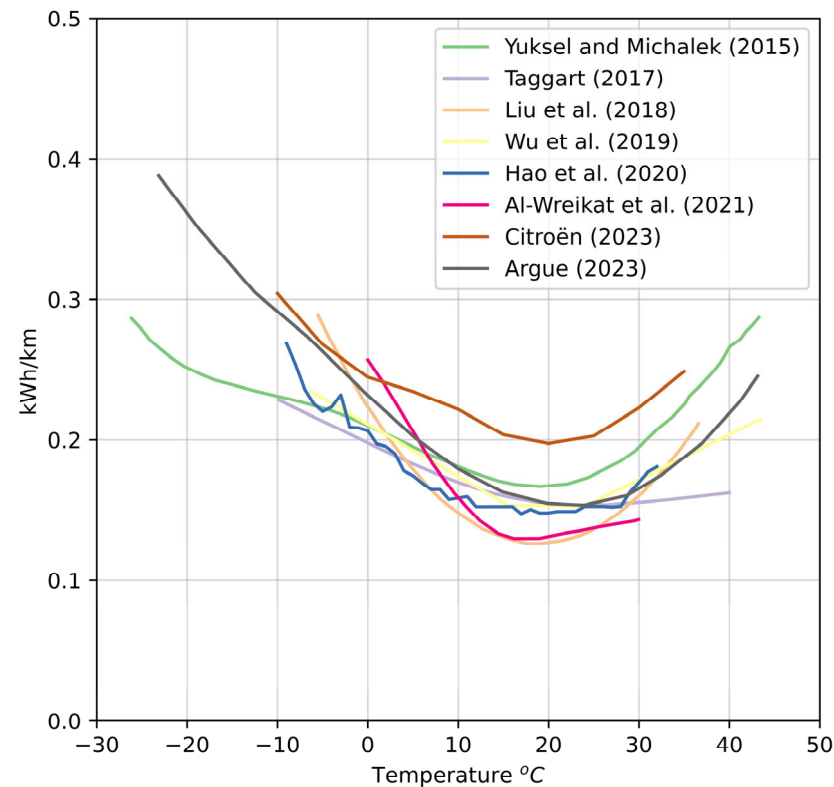
- **Yuksel & Michalek (2015)** *Effects of Regional Temperature on Electric Vehicle Efficiency, Range, and Emissions in the United States; and 7 additional sources.*

2. Fleet demand as a function of **multiple weather variables**

- **National Grid & Element Energy (2019)** *EV charging data; 8 million UK charging events in 2017.*

1. Fleet demand as a function of ambient temperature

We standardise these relationships of EV energy consumption and temperature using **changepoint regression**

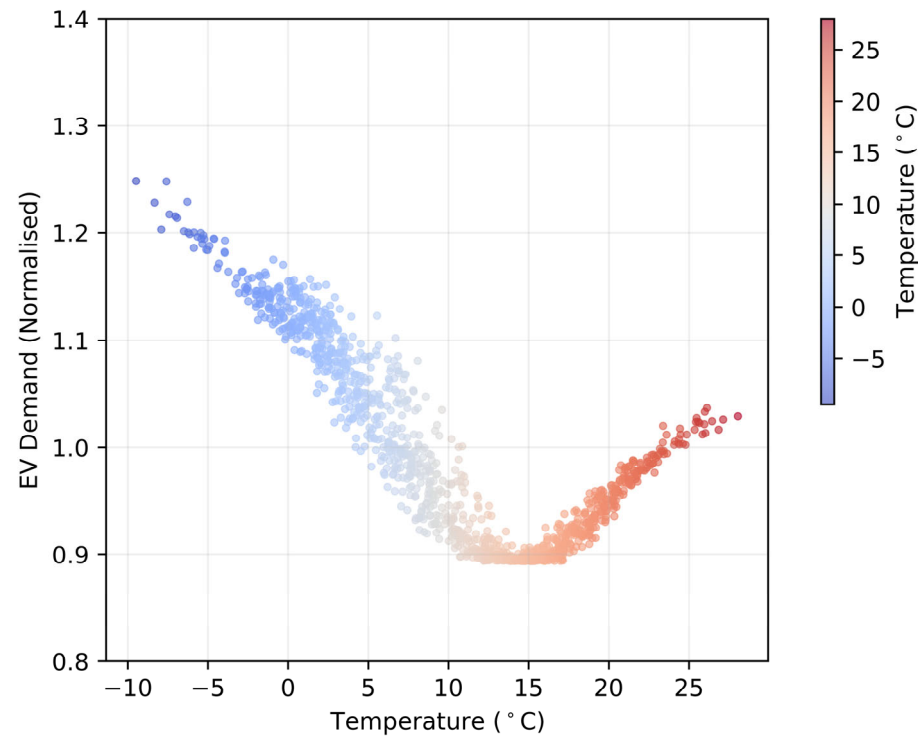


Below 15 °C or above 25 °C **energy consumption consistently increases**

These relationships with temperature are often **not included in studies of national-scale charging demand**

1. Fleet demand as a function of ambient temperature

We standardise these relationships of EV energy consumption and temperature using **change point regression**.



We use the *www.demand.ninja* model to **generate national-scale estimates of EV charging demand** covering many historical weather years.

2. Fleet demand as a function of multiple weather variables

- We use the National Grid data to **correlate fleet-wide EV consumption to other weather variables:**

	Estimate	Std. Error	t value	Pr(> t)
Intercept	45.270	0.595	76.050	(***) < 2e-16
<i>Temperature</i>	-0.513	0.076	-6.769	(***) 5.51e-11
<i>Solar Irradiance</i>	-0.014	0.002	0.795	(***) < 2e-16
Wind Speed	0.040	0.051	-0.772	0.441
Humidity	-0.116	0.151	-9.322	0.427
Weekend	-9.727	0.189	-51.470	(***) < 2 e-16

Residual standard error: 1.622 on 348 degrees of freedom;

Multiple R-squared: 0.9258, adjusted R-squared: 0.9248, F-statistic: 8686.6 on 5 and 348 DF, p-value: <2.2e-16.

2. Fleet demand as a function of multiple weather variables

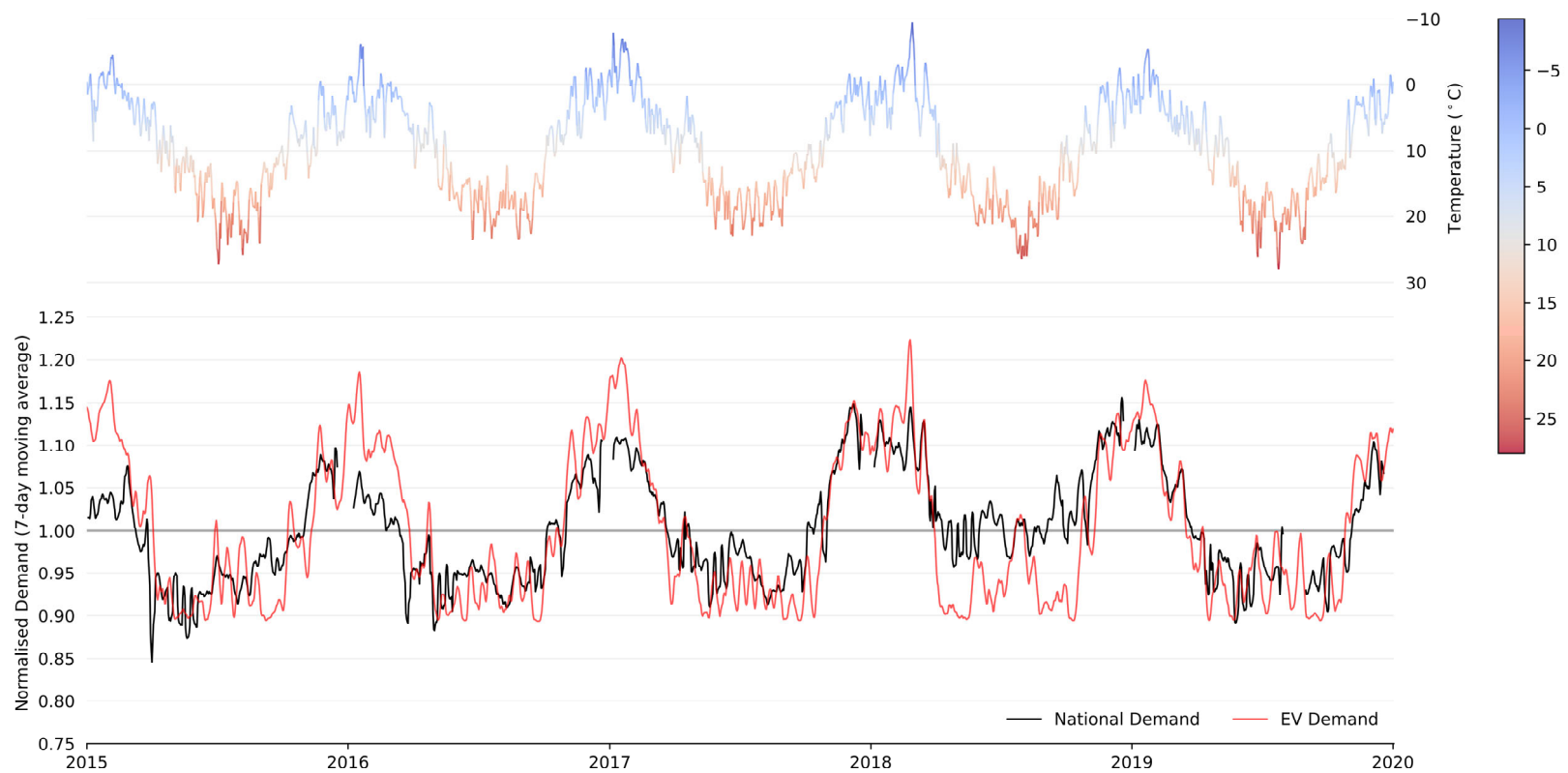
- We use the National Grid data to **correlate fleet-wide EV consumption to other weather variables:**
 - Intercept: Baseline demand of 45.3 MW
 - Weekends: On weekends demand drops by 21.4%
 - Temperature: Demand increases by 1.1% for every 1 °C drop
 - Solar Irradiance: Demand increases by 3.1% for every 100 W/m²* drop

*(annual UK range is ± 300 W/m²)

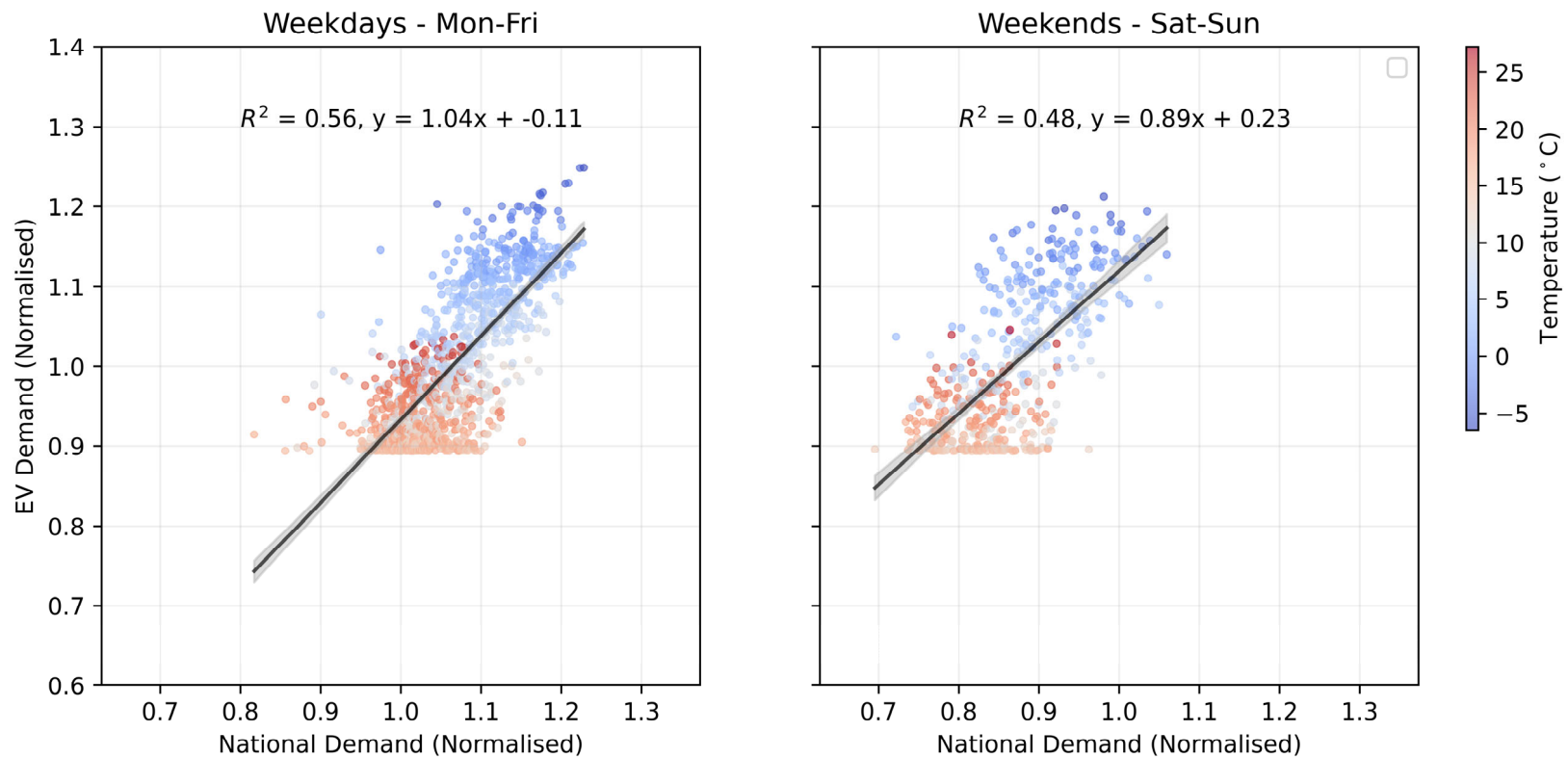
EV demand can be modelled as a function of the weather

1. Fleet demand as a function of **ambient temperature** -> **German case study**
 2. Fleet demand as a function of **multiple weather variables** -> **UK case study**
- We use the *www.demand.ninja* model to **generate national-scale estimates of EV charging** demand covering many historical weather years.
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Case study: German fleet

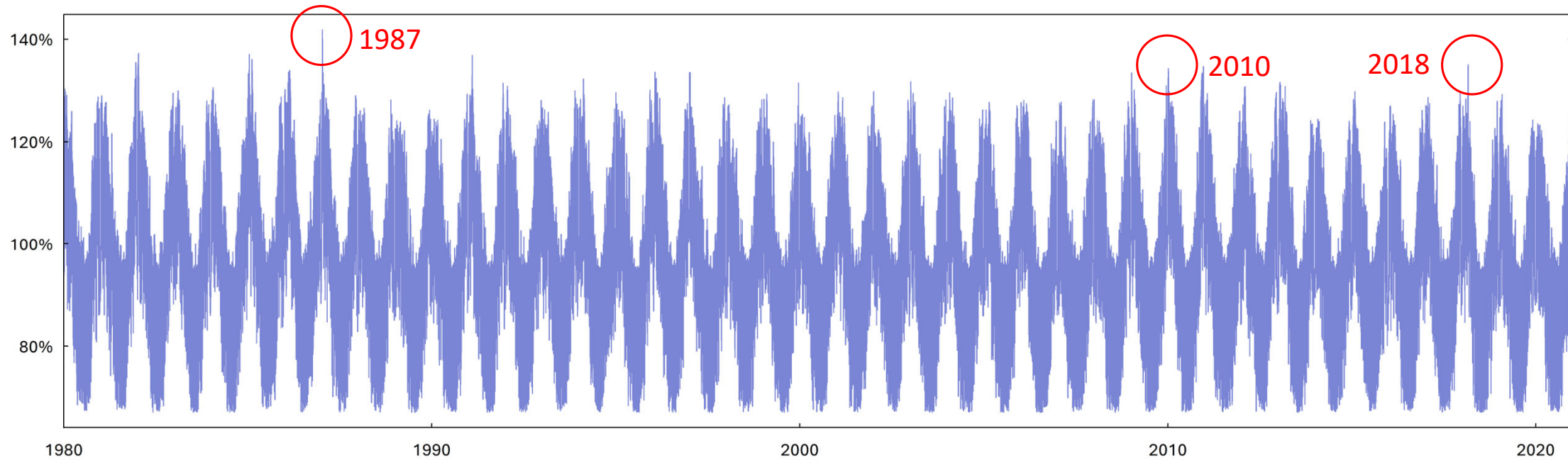


Case study: German fleet



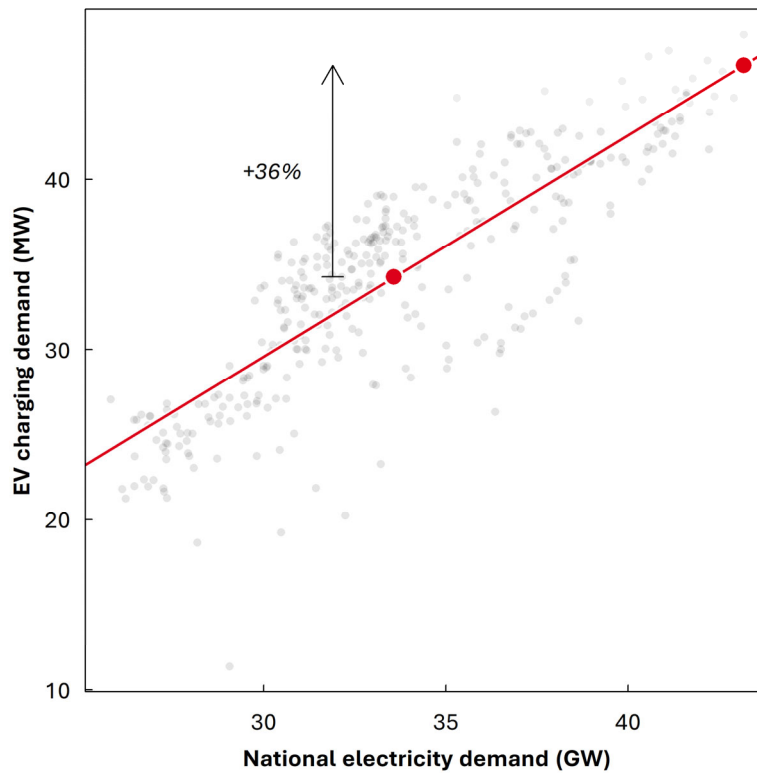
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Case study: UK Fleet



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Case study: UK Fleet



Charging demand was **36% higher on the day of peak electricity demand** than the annual-mean.

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lain could you confirm if it was 36% higher than the annual mean for EVs or National demand?

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Case studies summary

- **EV demand varied seasonally**, on average increasing by 12% in winter and decreasing by 6% in summer for Germany, compared to the annual average.
 - During highest day of peak demand, **EV demand increased by 25% and 36%** compared to the annual average for Germany and the UK, respectively.
 - **Correlation between EV demand and national demand**, suggesting demand from **EVs will contribute to peak conditions**.
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Conclusions

- Evidence **EV energy consumption is impacted by the weather**, therefore future models should include this relationship.
 - EV fleet demand will follow **seasonal, weekly, and daily, weather patterns** that will need to be balanced with corresponding supply.
 - Demand from future EV fleets is likely to **contribute to peak conditions**.
 - Likely to **increase system cost** and if not addressed **may risk security of supply**.
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Next Steps

- Global modelling – apply these methods to different regions.
 - EV Ninja – online webtool to generate EV charging profiles for different locations.
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Thank you

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