PV and opportunistic electric vehicle charging in a Swedish distribution grid

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Introduction

What we study

How we do it



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- 10kV / 400V three-phase power grid
- 5174 grid nodes / end-users
- Only electric vehicles (EV) in the car fleet
- Over- and undervoltage due to
 - High load (mainly winter)
 - High PV generation (mainly summer)

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 - High PV generation (mainly summer)

- PV potential using LiDAR data
- PV penetration 0-100% of yearly load
- Markov-chain EV charging model
- Newton-Raphson power flow solution



PV generation & load data

 Rooftop PV power potential using GIS, LiDAR and irradiance data





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 - Hourly load for 5174 end-users (2014)





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- Rooftop PV power potential using GIS, LiDAR and irradiance data
- DSO 'Herrljunga Elektriska'
 - Hourly load for 5174 end-users (2014)
- Yearly PV penetration with randomly selected rooftops
 - 0%
 - 10%
 - ...
 - 90%
 - 100%





- 2 MV grids
- 338 LV grids (rural & city)
- 3891 nodes, 5174 end-uses





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- Hourly load data
- Allowed end-user voltage
 - Max 1.1 pu
 - Min 0.9 pu
- Always 1.0 pu at the primary substations









 Opportunistic EV charging – charging whenever & wherever parked

For more information: M. Shepero and J. Munkhammar. *Modelling charging of electric vehicles using mixture of user behaviours*. 1st E-Mobility Integration Symposium, October 23rd, Berlin



- Opportunistic EV charging charging whenever & wherever parked
- Time dependent (time of day, weekend/weekday)

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- Markov chain with 3 states
 - Home,
 - Work
 - Other (public parking lots)





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- Markov chain with 3 states
 - Home
 - Work
 - Other (public parking lots)
- 2 summer + 2 winter weeks
- Charging power: 3.7 kW





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 - 5295 vehicles in 2016 in the municipality
 - 333 extra EVs in the summer (summer houses)



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- Aggregated 1-minute EV charging data to hourly resolution

$$E_t^n = \begin{cases} E_{t-1}^n + 3.7 \times \Delta t & \text{if charging,} \\ E_{t-1}^n - \eta \times D & \text{if driving,} \\ E_{t-1}^n & \text{else,} \end{cases}$$



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3.7 kW charging power × time Battery charge per EV at time t $E_t^n = \begin{cases} E_{t-1}^n + 3.7 \times \Delta t & \text{if charging,} \\ E_{t-1}^n - \eta \times D & \text{if driving,} \\ E_{t-1}^n & \text{else,} \end{cases}$

Consumption per km × driving distance (km)



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- Small difference in load with EV
 - 18% higher in the summer weeks
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- Large seasonal variation in PV generation
 - 100% penetration in the figures on a yearly basis









Customers with undervoltage [%]

Undervoltage [customer-hours]

1.5

0.5

Results – undervoltage Customers with undervoltage [%] 1.5 Summer Winter 0.5 C Undervoltage [customer-hours] Summer Winter

Yearly PV penetration [%]

Yearly PV penetration [%]





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- 50% of the customers are affected by overvoltage in a scenario of 100% PV penetration – almost no reduction with EV charging
 - Overvoltage in LV grids far from the distribution substations
 - EV charging during day mainly in the city areas close to substations



 PV power has a small impact on undervoltage due to EV charging in the winter, in the summer with PV > 50%



- PV power has a small impact on undervoltage due to EV charging in the winter, in the summer with PV > 50%
- 1.5% of the customers affected by undervoltage in the winter
 - Undervoltage in LV grids far from the distribution substations
 - EV charging mainly in the morning (to work) and in the afternoon (to home)
 - Sun is above the horizon approx. 08:40 15:30 in early January



- Possible solutions to avoid voltage limit violations
 - Grid extension can be costly for rural grids
 - 'Smart-grid', for example real-time measurements with tapchanging transformers
 - Scheduled EV charging or 'vehicle to grid' incentives are needed



Thank you for listening!

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