

# Grid Load Peak Prediction with sonnen Final Presentation

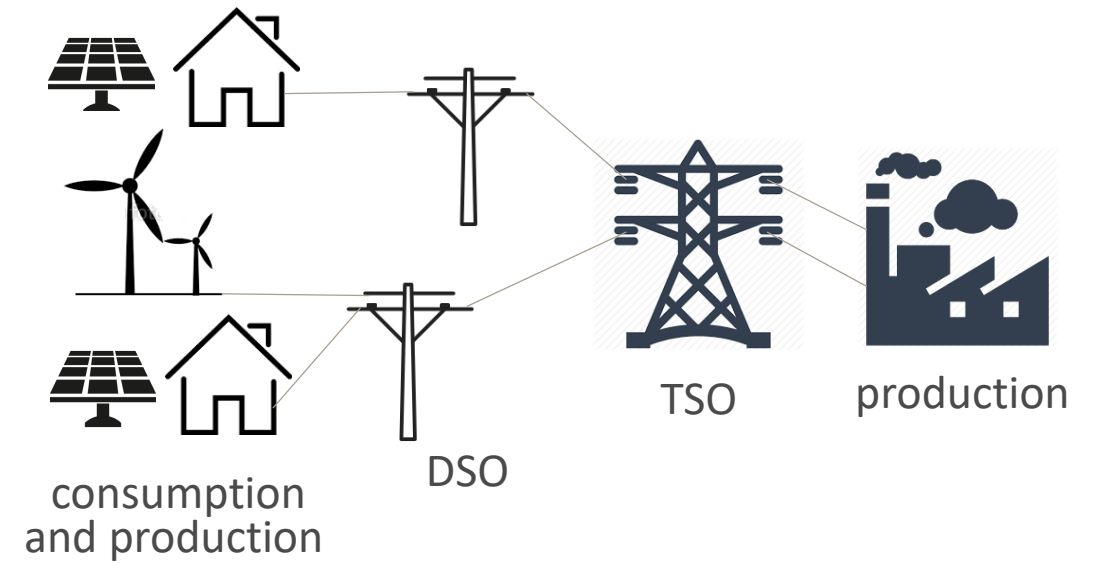
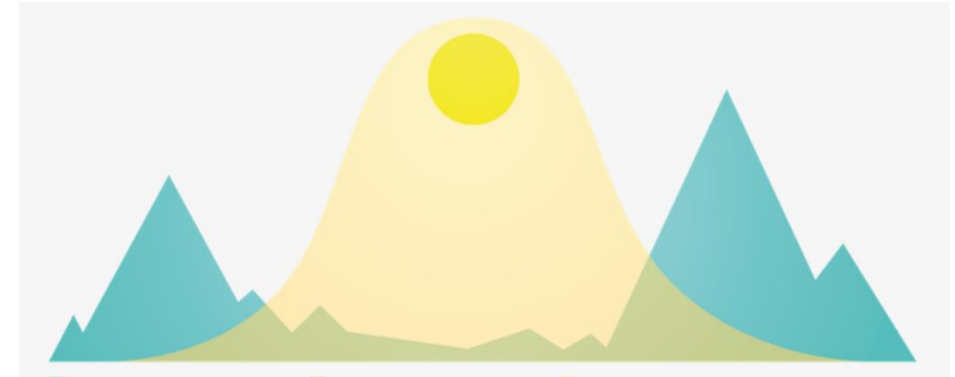
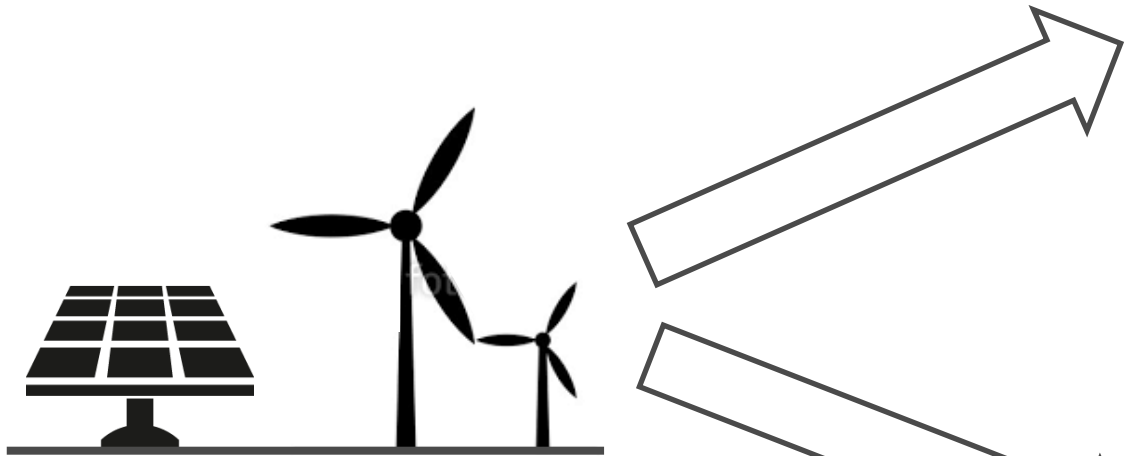
TUM Data Innovation Lab

Constantin Gahr, Ayishetu Haruna and Chris Löchel

# Agenda

- 1) Problem description
- 2) Data exploration
  - Power data
  - Weather data
- 3) Models
  - Statistical model
  - SVM model and neural network
  - ARIMA model
- 4) Web application
- 5) Conclusion

# Problem description



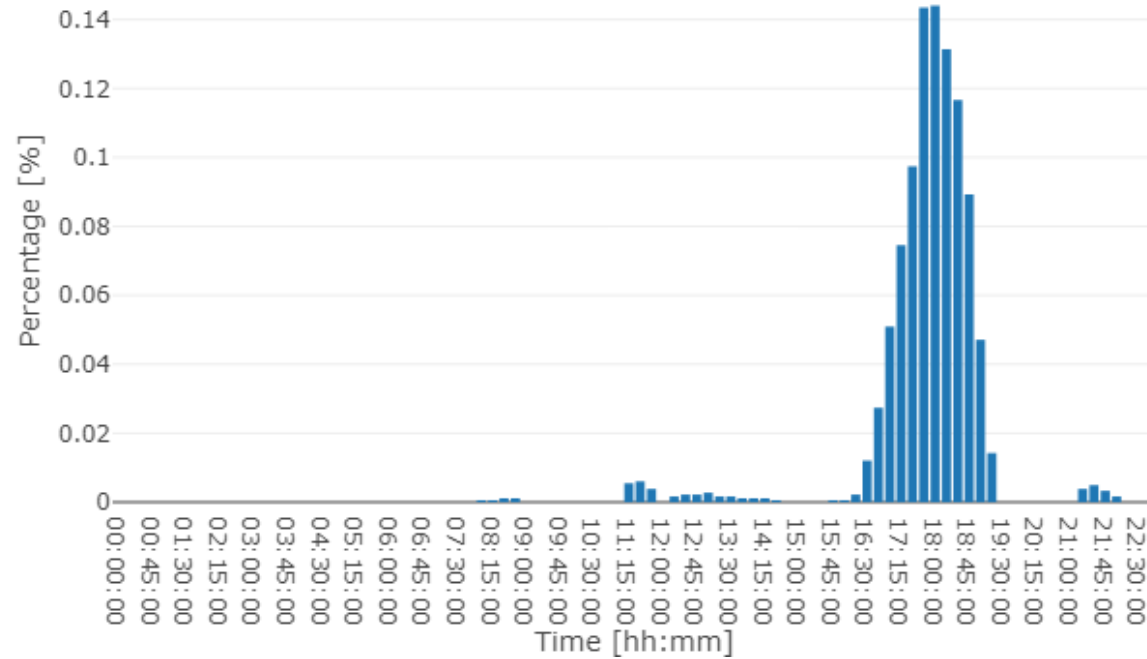
# Data collection process



→ DSOs must publish grid load of the past year

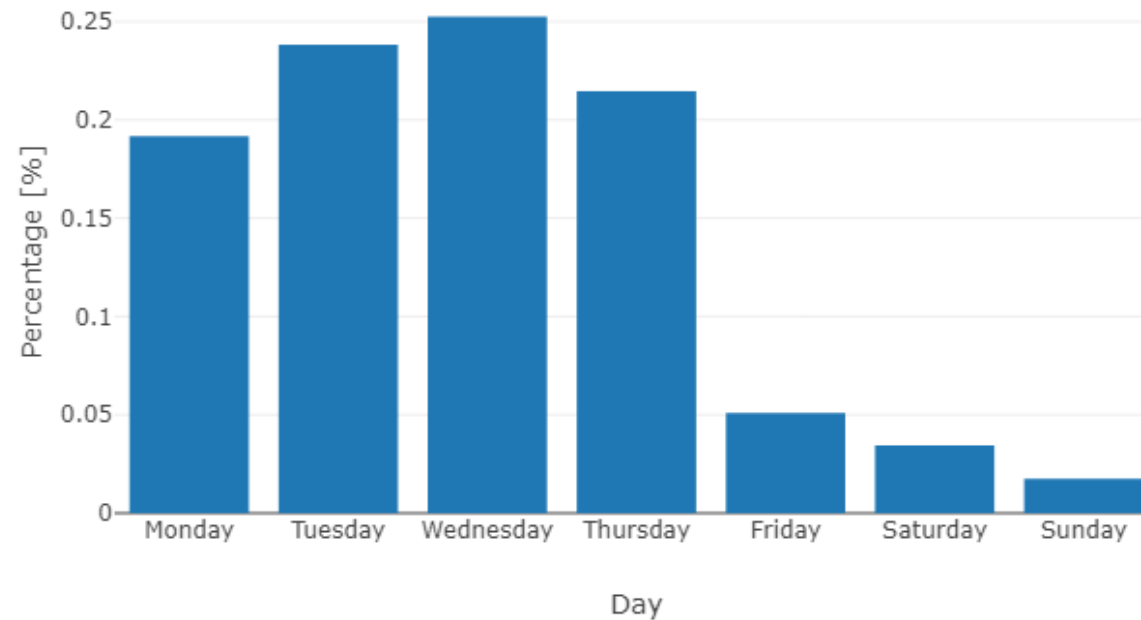
# Statistical data exploration

Distribution of power usage within a day, values higher than 0.95 \* peak value



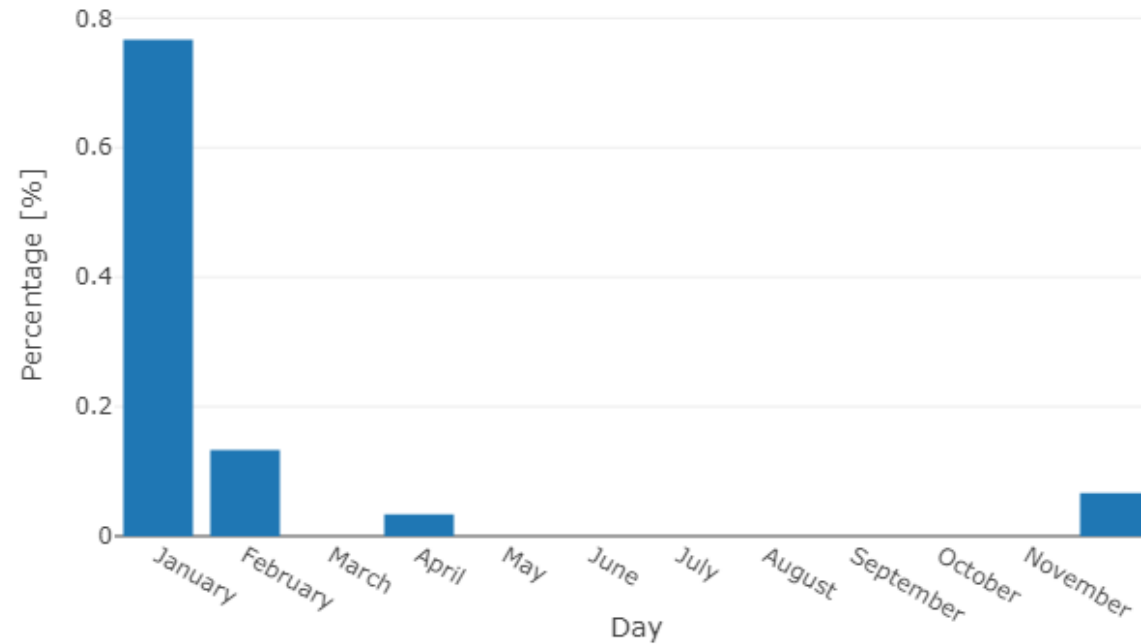
# Statistical data exploration

Distribution of power usage within a week, values higher than  $0.95 \cdot \text{peak value}$



# Statistical data exploration

Distribution of power usage within a year, values higher than 0.95 \* peak value

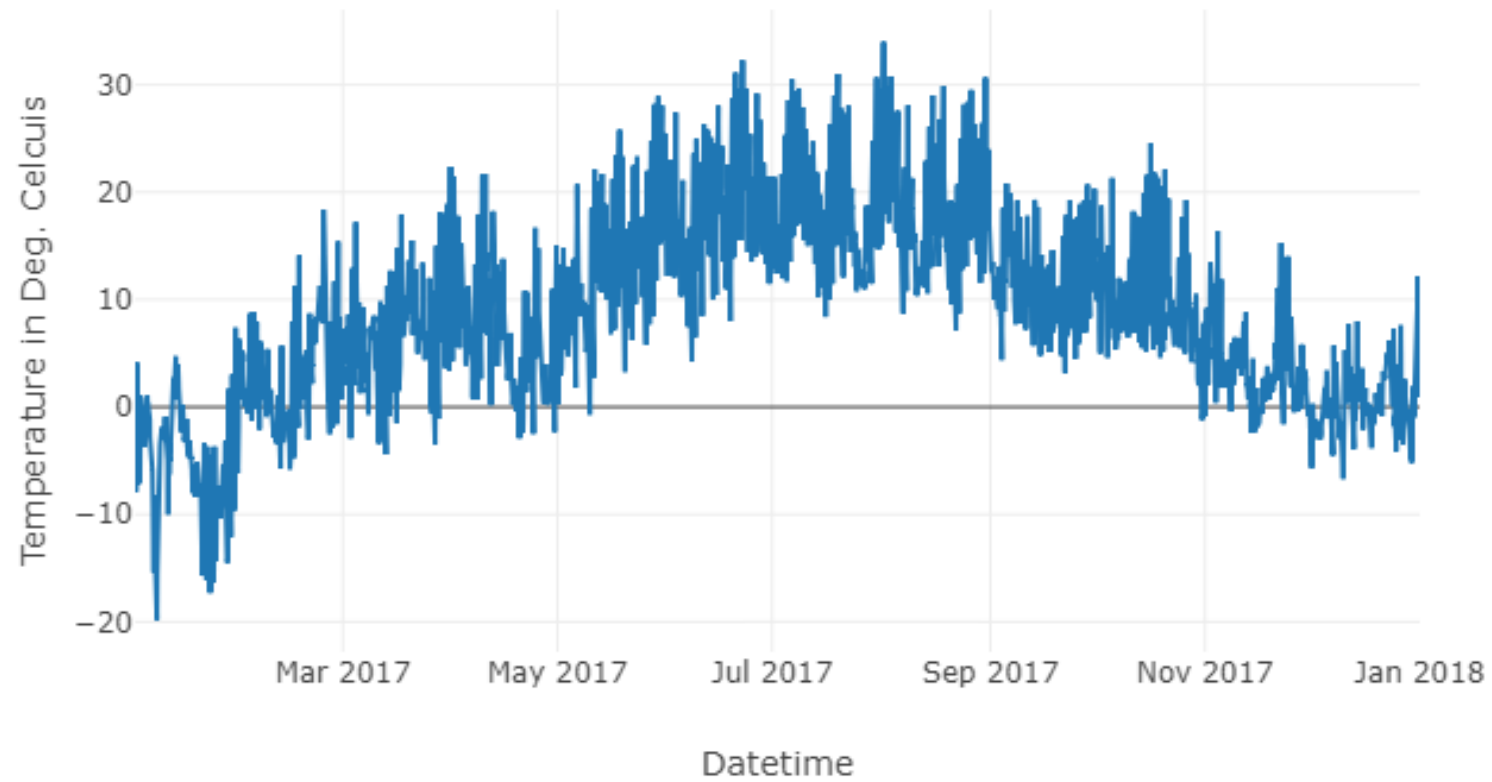


# Weather Data

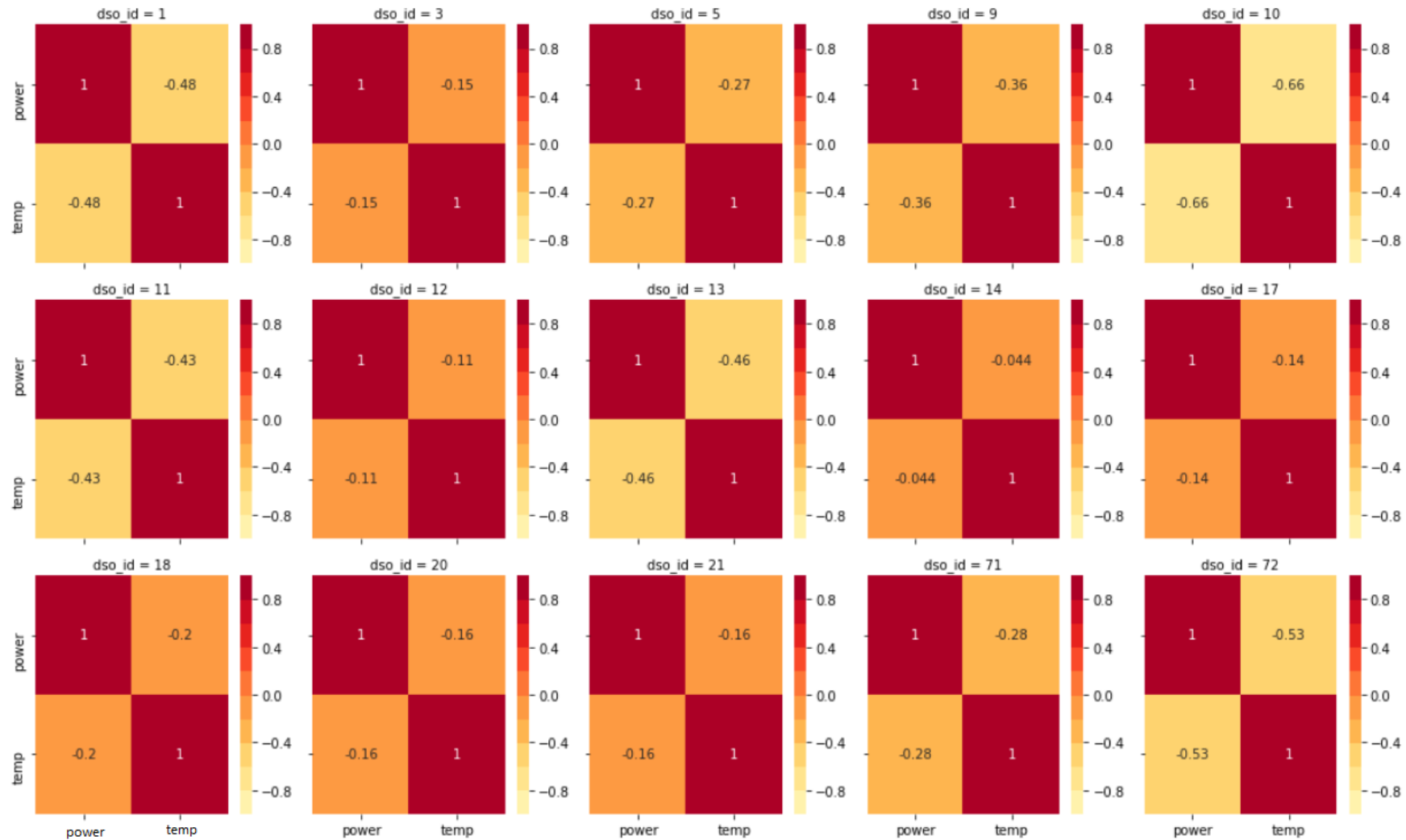
- Climate Data Center (Deutscher Wetterdienst)
- Hourly temperature observations in degree celsius



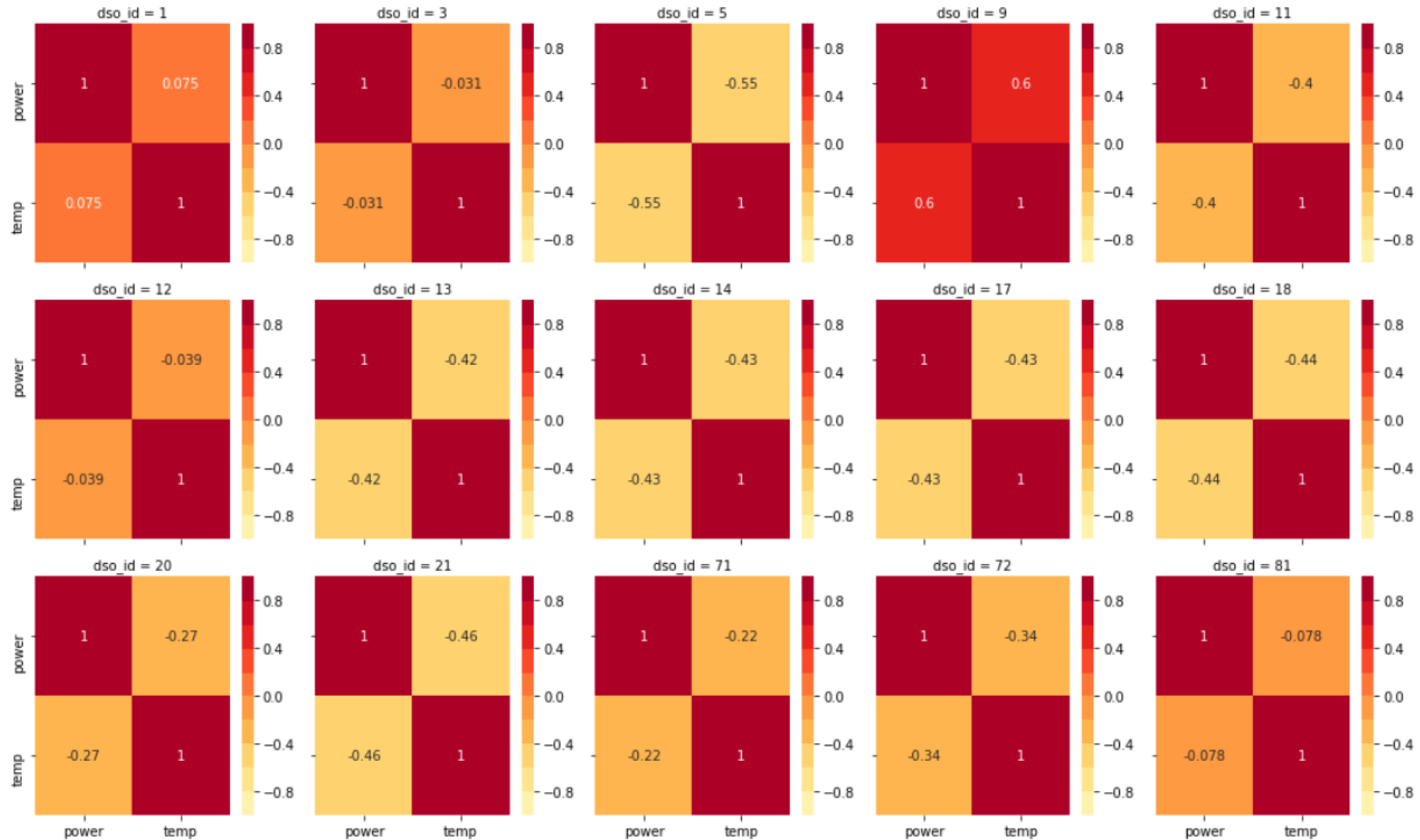
# Weather Data



# Correlation: Power and Temperature



# Correlation: Power (Highest) and Temperature

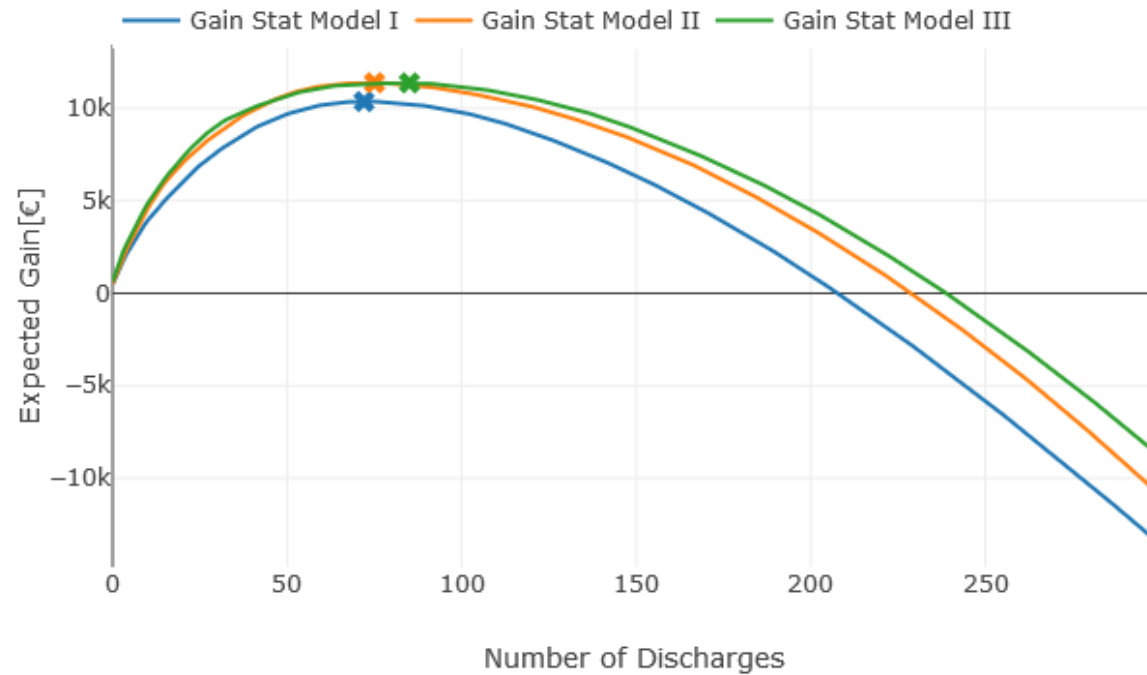


# Statistical Model

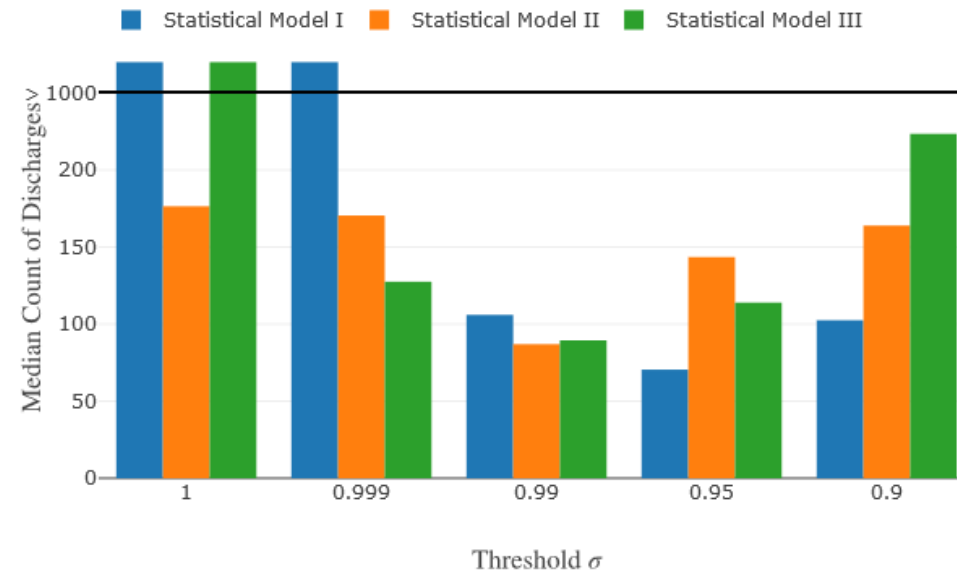
Idea:

$$\begin{aligned} P(\mathcal{P}(t, d, m) \geq \sigma) &= P(\{T_\sigma = t\} \cap \{\tilde{D}_\sigma = d\} \cap \{M_\sigma = m\}) \\ &= \rho_{T,\sigma}(t) \cdot \rho_{\tilde{D},\sigma}(d) \cdot \rho_{M,\sigma}(m) \\ &\approx \tilde{\rho}_{T,\sigma}(t) \cdot \tilde{\rho}_{\tilde{D},\sigma}(d) \cdot \tilde{\rho}_{M,\sigma}(m). \end{aligned}$$

# Statistical Model – Theoretical Performance



# Statistical Model – Practical Performance



Performance statistical model

Gain: ca. 20000€ per MW

# SVM and NN – Data Preparation

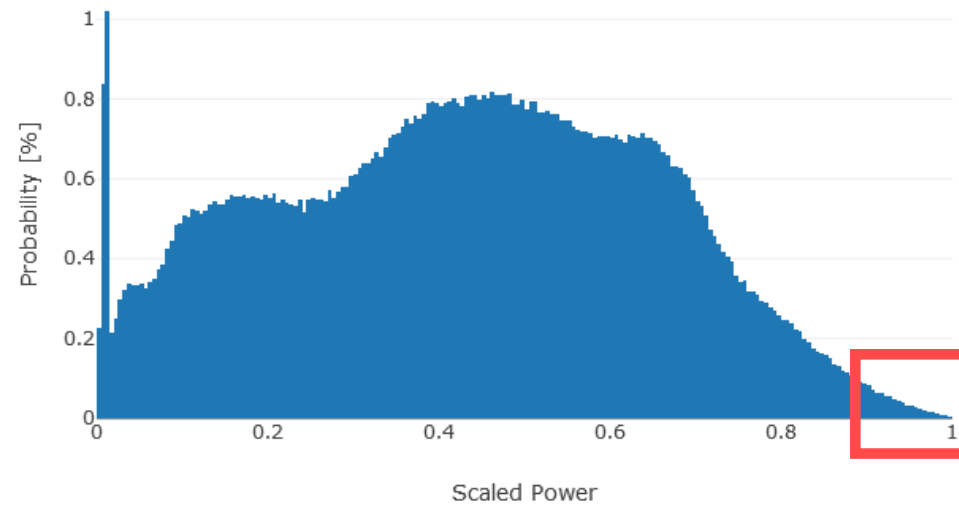
Idea: Create NN and SVM which uses weather data

## Data Preparation:

- One Hot Encoding for datetime data
- Temperature data as continuous data

# SVM and NN – Data Preparation

But: Dataset is highly unbalanced -> subsampling





# SVM and NN – Configuration

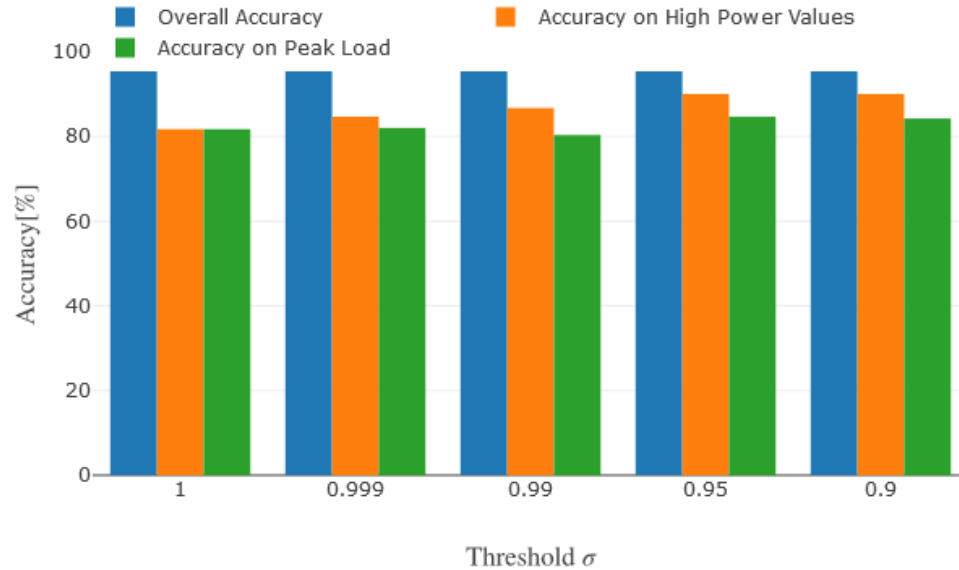
## Support Vector Machine:

- Gaussian Kernel

## Neural Network:

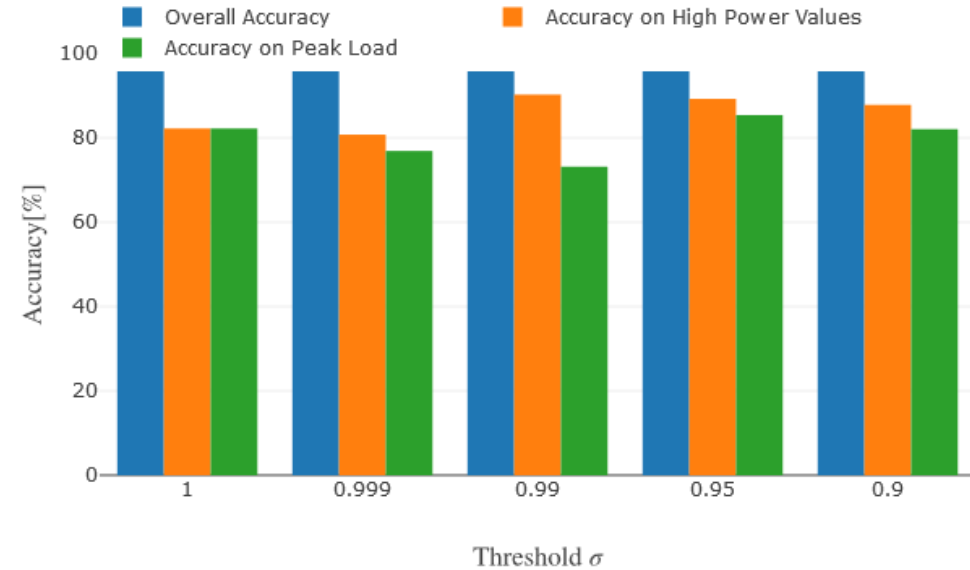
- 3 layers (12 – 8 – 1 neurons)
- 48 inputs (temperature and one hot encoded date)
- Relu activation function  $f(x) = \max(0, x)$  in first two layers
- Sigmoid in last layer

# NN and SVM – Performance



Performance support vector machine

Gain: ca. 47000€ per MW



Performance neural network

Gain: ca. 38000€ per MW

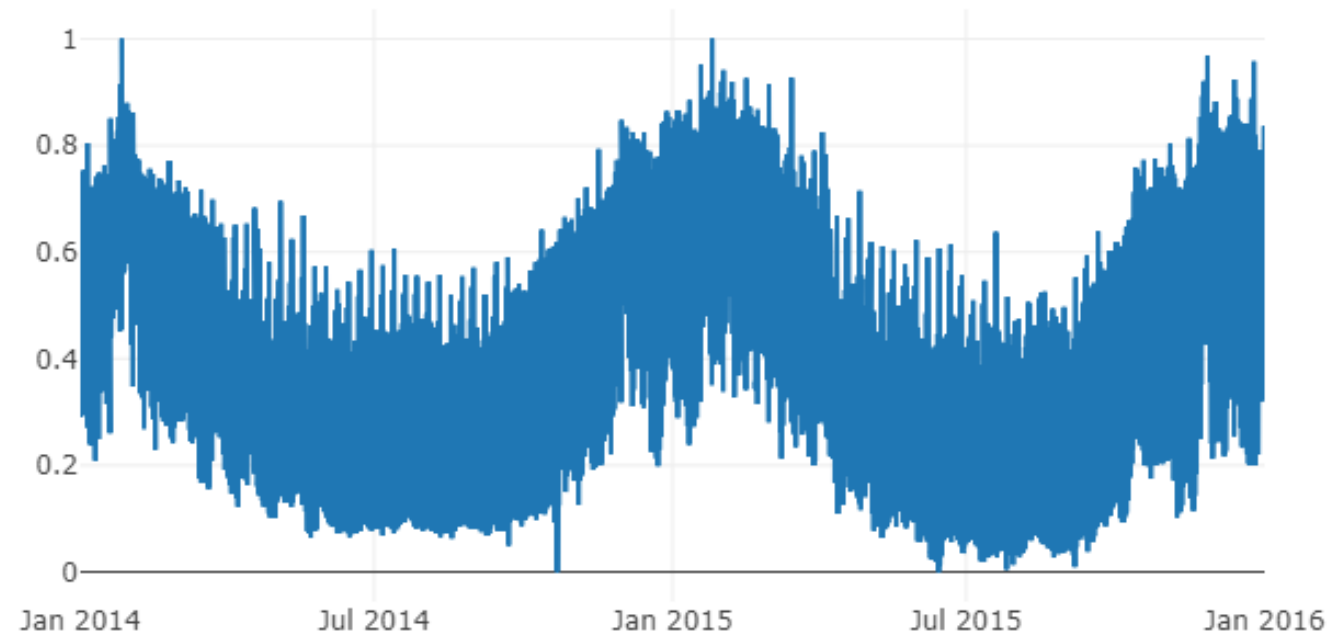
# ARIMA model

ARIMA model = Autoregressive Integrated Moving Average model

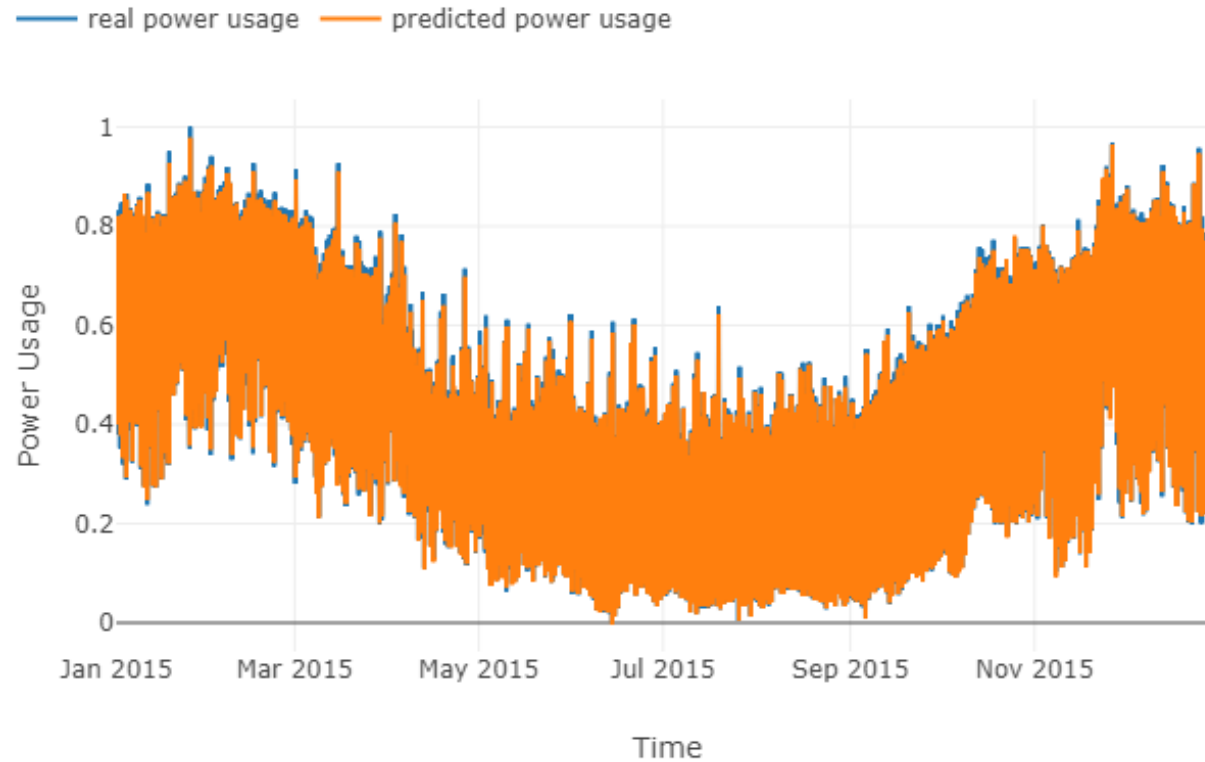
- Needs **stationary** data
  - Integrated into Python library statsmodels
- 
- STEP 1: Make data stationary
    - Only predict difference between current and next value
    - *Prediction for next value = average of the past two values + predicted difference*
  - STEP 2: Choose parameters
  - STEP 3: Fit ARIMA model and run prediction

# ARIMA model

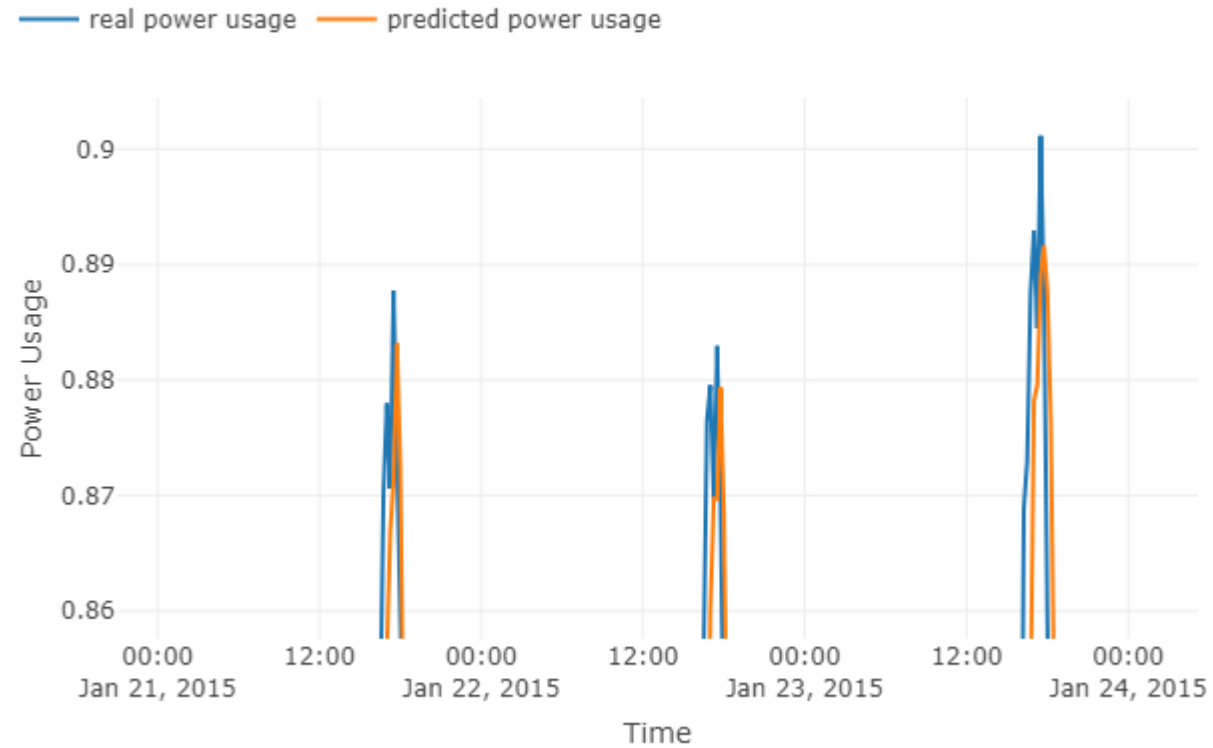
Avacon Netz GmbH, years of 2014 and 2015



# ARIMA model prediction



# ARIMA model prediction



# Choosing peak values of ARIMA prediction

Initialize `current_max` and choose threshold `T`

## For every time intervall

if prediction for next interval  $>$  `current_max` \* `T`:

    update `current_max`

    discharge batteries

# Evaluation of ARIMA model

- Choose different values for T
  - How often is electricity fed into the grid?
  - Is the maximum identified?
- Compute gain/loss



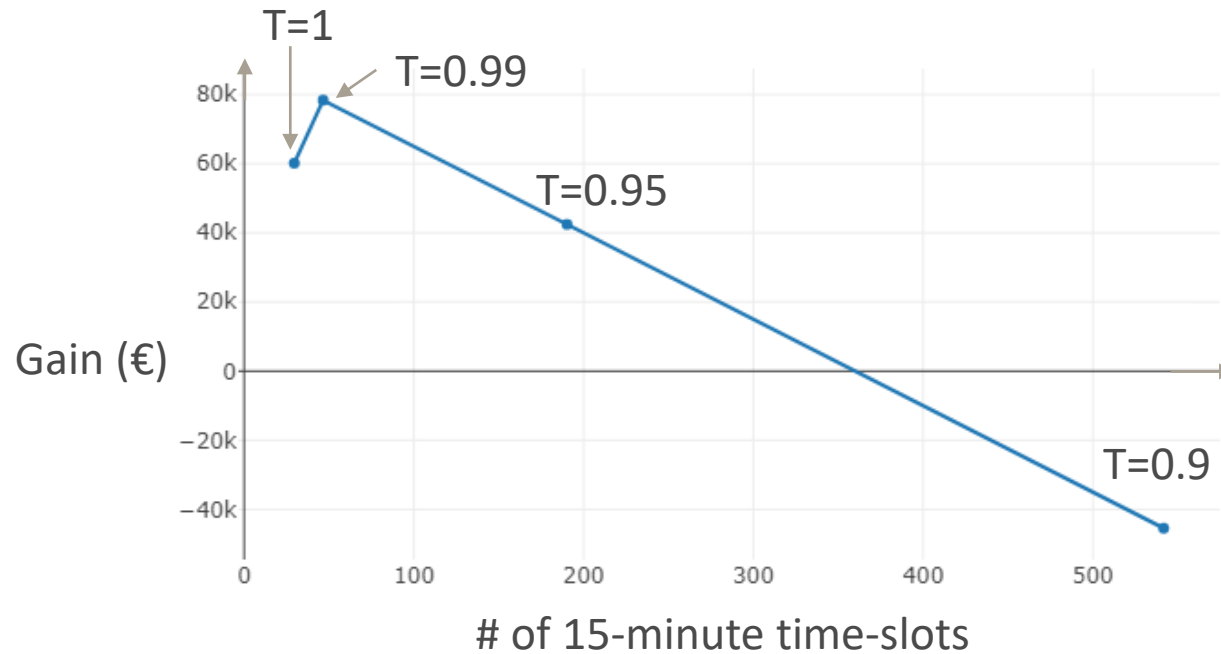
# Evaluation of ARIMA model

## Performance for four DSOs:

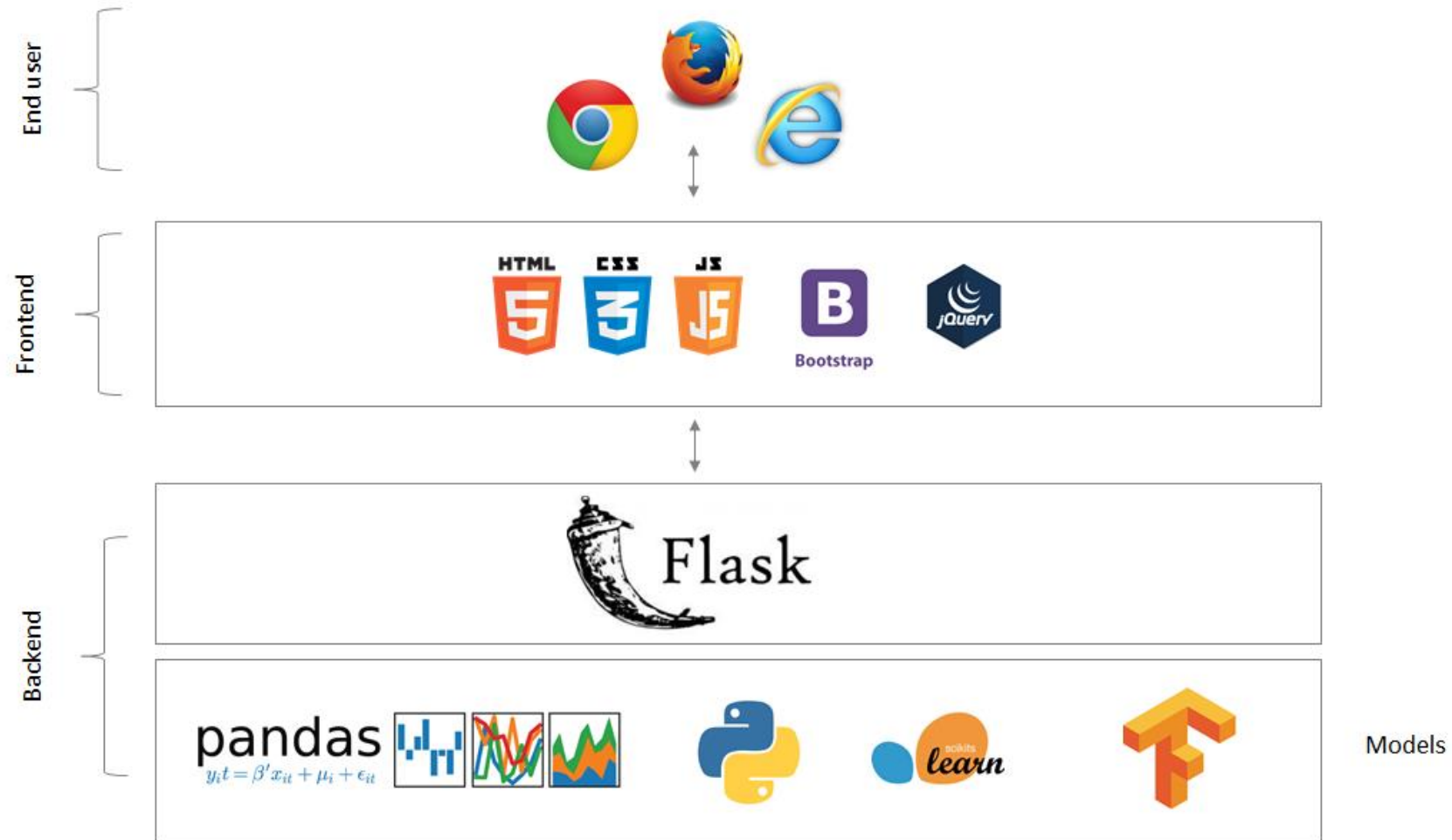
DSO ID		T				
		1	0.99	0.95	0.9	0.85
71	# of feed-ins	17	21	167	579	1194
	maximum hit?	1	1	1	1	1
	gain/loss	85750	84750	48250	-54750	-208500
72	# of feed-ins	23	33	76	270	600
	maximum hit?	1	1	1	1	1
	gain/loss	84250	81750	71000	22500	-60000
12	# of feed-ins	38	61	162	404	794
	maximum hit?	1	1	1	1	1
	gain/loss	80500	74750	49500	-11000	-108500
14	# of feed-ins	40	71	356	914	1945
	maximum hit?	0	1	1	1	1
	gain/loss	-10000	72250	1000	-138500	-396250
Average	# of feed-ins	29,5	46,5	190,25	541,75	1133,25
	maximum hit?	0,75	1	1	1	1
	gain/loss	60125	78375	42437,5	-45437,5	-193312,5

# Evaluation of ARIMA model

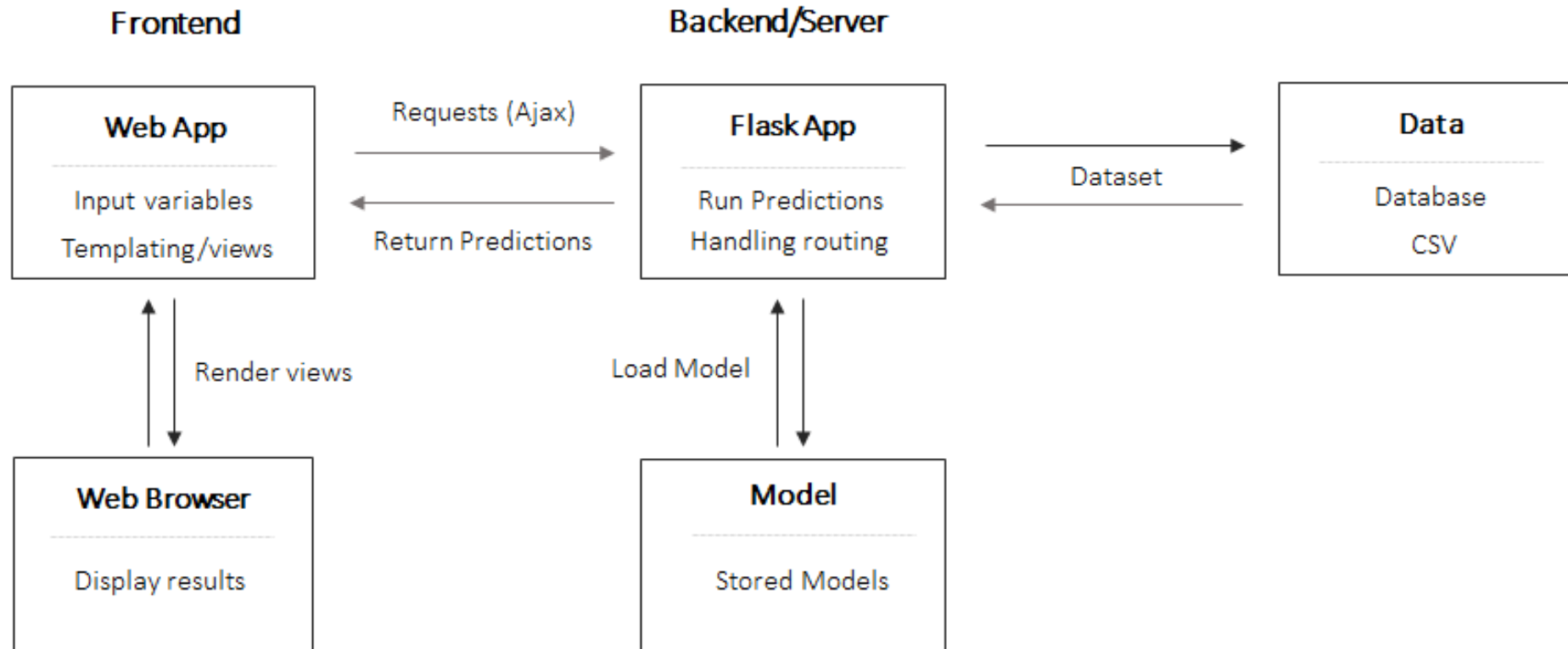
## Performance curve for four DSOs:



# Web Application: Architecture



# Web Application: Application Process flow



# Conclusion

- Statistical Model:
  - Gain: ca 20000€ per MW
  - Baseline performance
  - Ready for production
- NN and SVM:
  - Gain: ca 40000€+ per MW
  - Room for improvement (optimize hyperparameters, use live data, ...)
- ARIMA Model:
  - Gain: ca 78000€ per MW (to be taken with a grain of salt!)
  - Needs live data
- Web Application:
  - Demonstration of our models

Thank you for your attention!