



ML-BASED PRICE FORECASTS IN THE OPEN ELECTRICITY MARKET MODEL AMIRIS

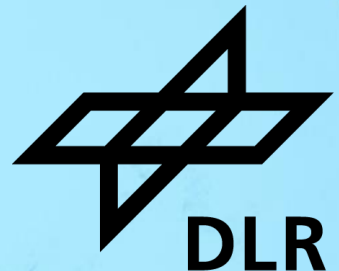
Felix Nitsch^{1,*}, Christoph Schimeczek¹

February, 2025

Acknowledgements: Valentin Bertsch, Kristina Nienhaus on behalf of the Energy Economics Group

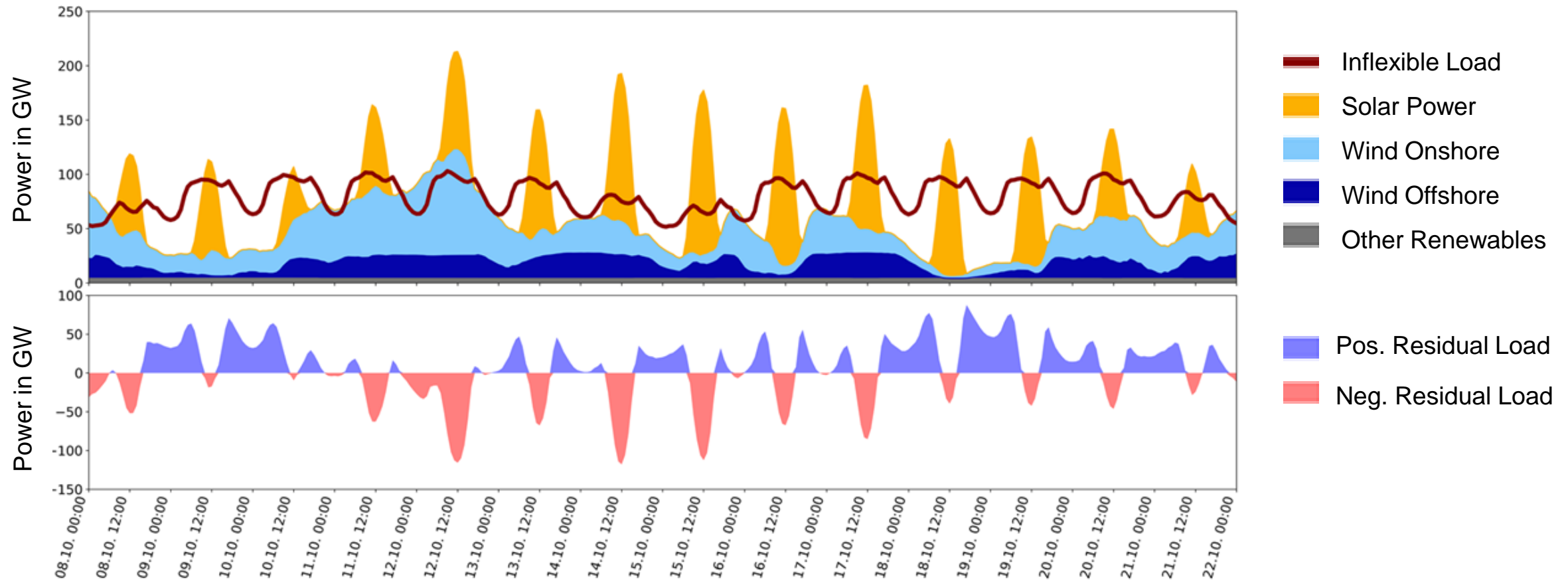
¹ German Aerospace Center | Institute of Networked Energy Systems | Energy Systems Analysis

* felix.nitsch@dlr.de



Introduction

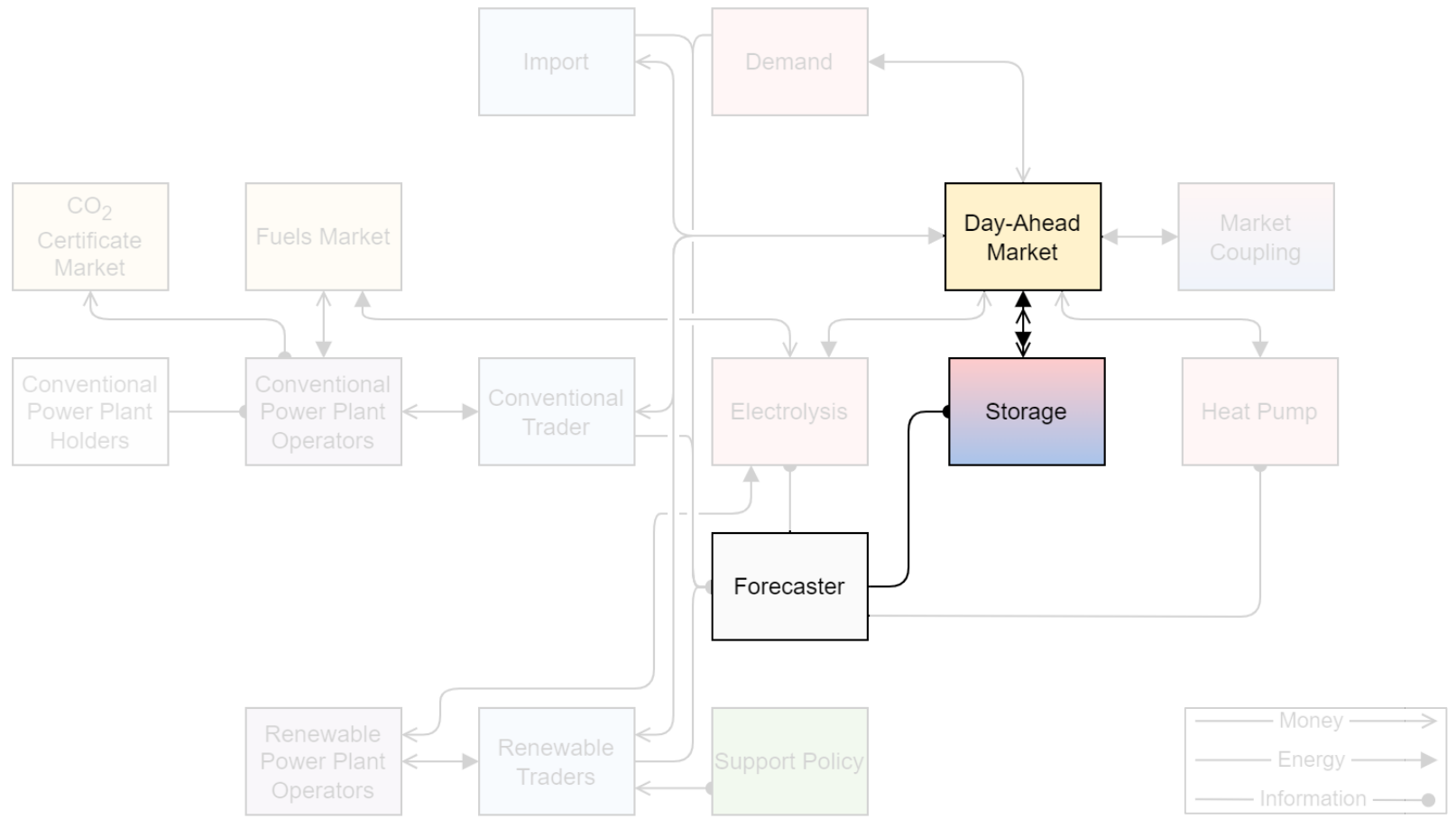
Challenges of Intermittent Energy Sources



Renewable energy generation in Germany with inflexible load (top) and resulting residual load (bottom) for two weeks in a scenario year 2045

AMIRIS

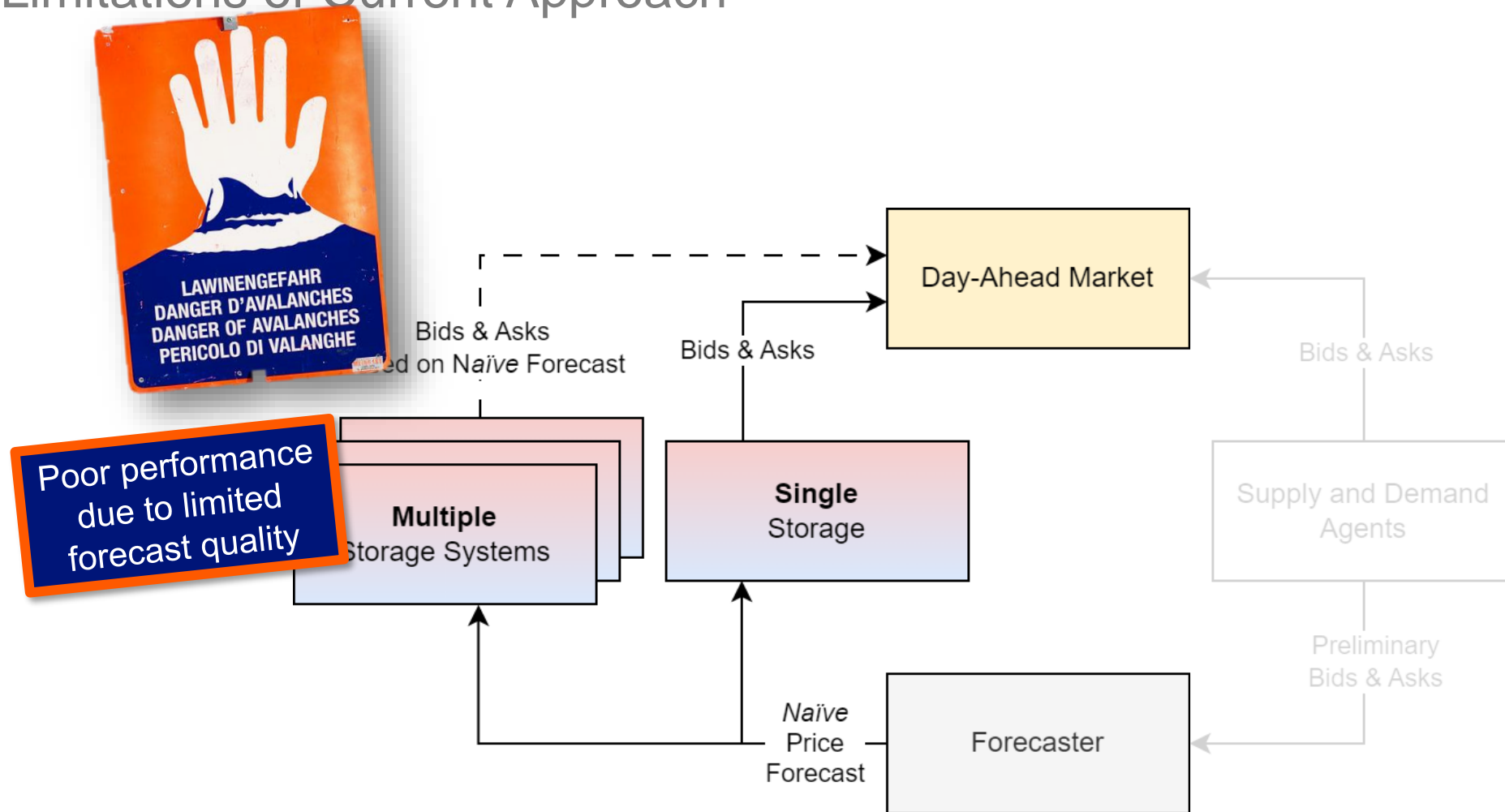
Open Agent-based Electricity Market Model



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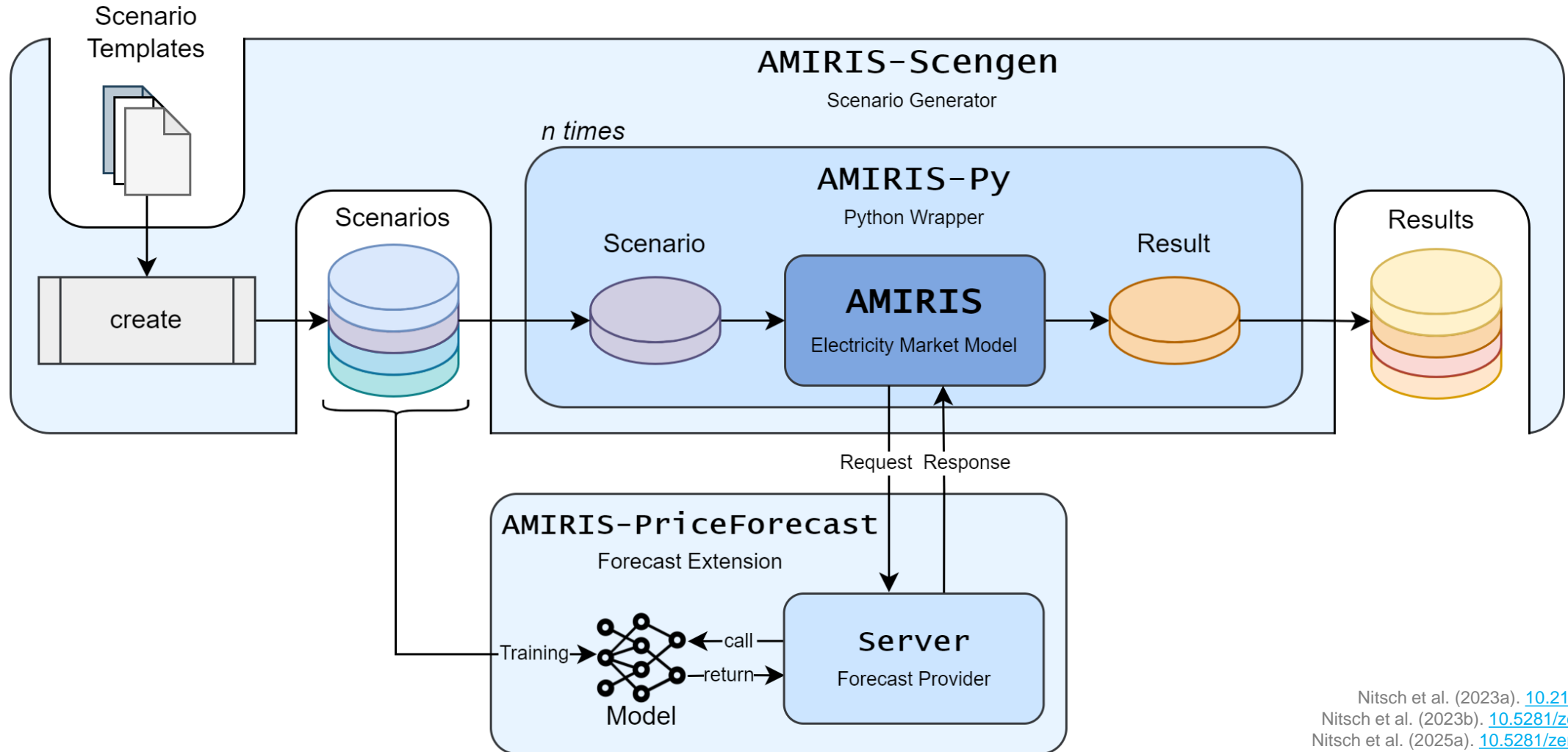
Modelling Flexibility Options in AMIRIS

Limitations of Current Approach



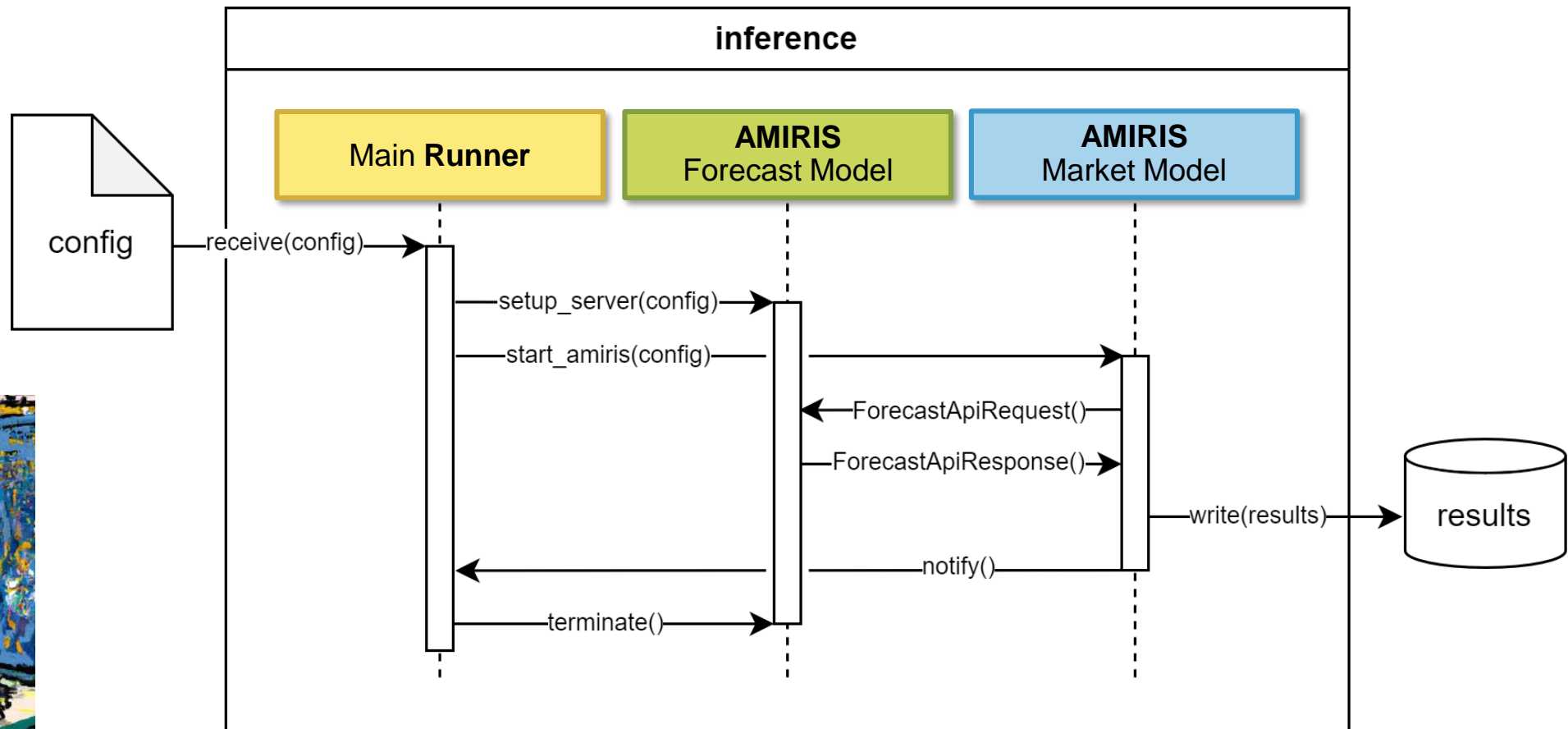
Model Development and Simulation Setup

Enabling Machine-Learning Based Electricity Price Forecasts



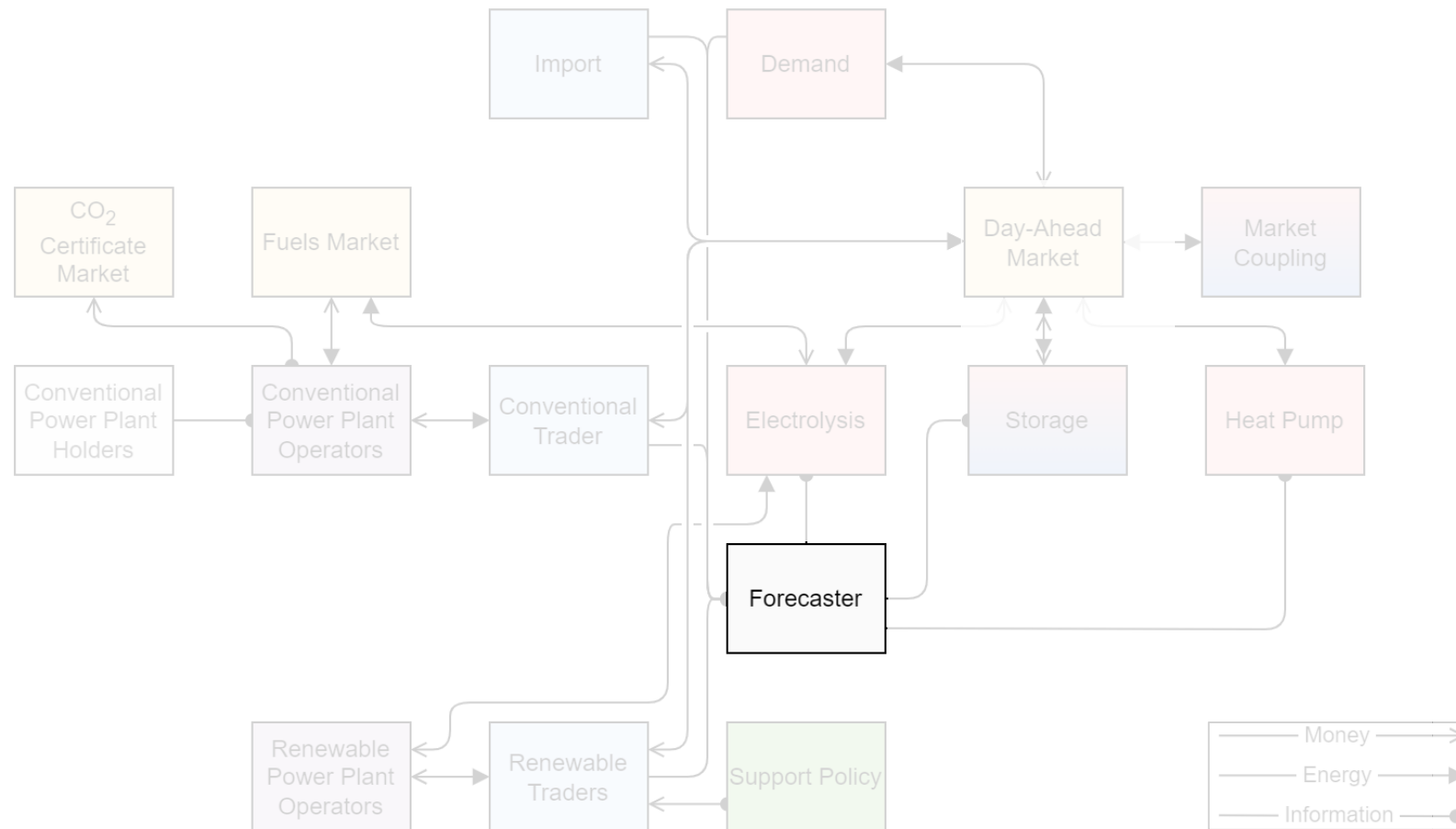
Connecting AMIRIS to External Forecasting Model

Coupling using FastAPI



AMIRIS

Open Agent-based Electricity Market Model



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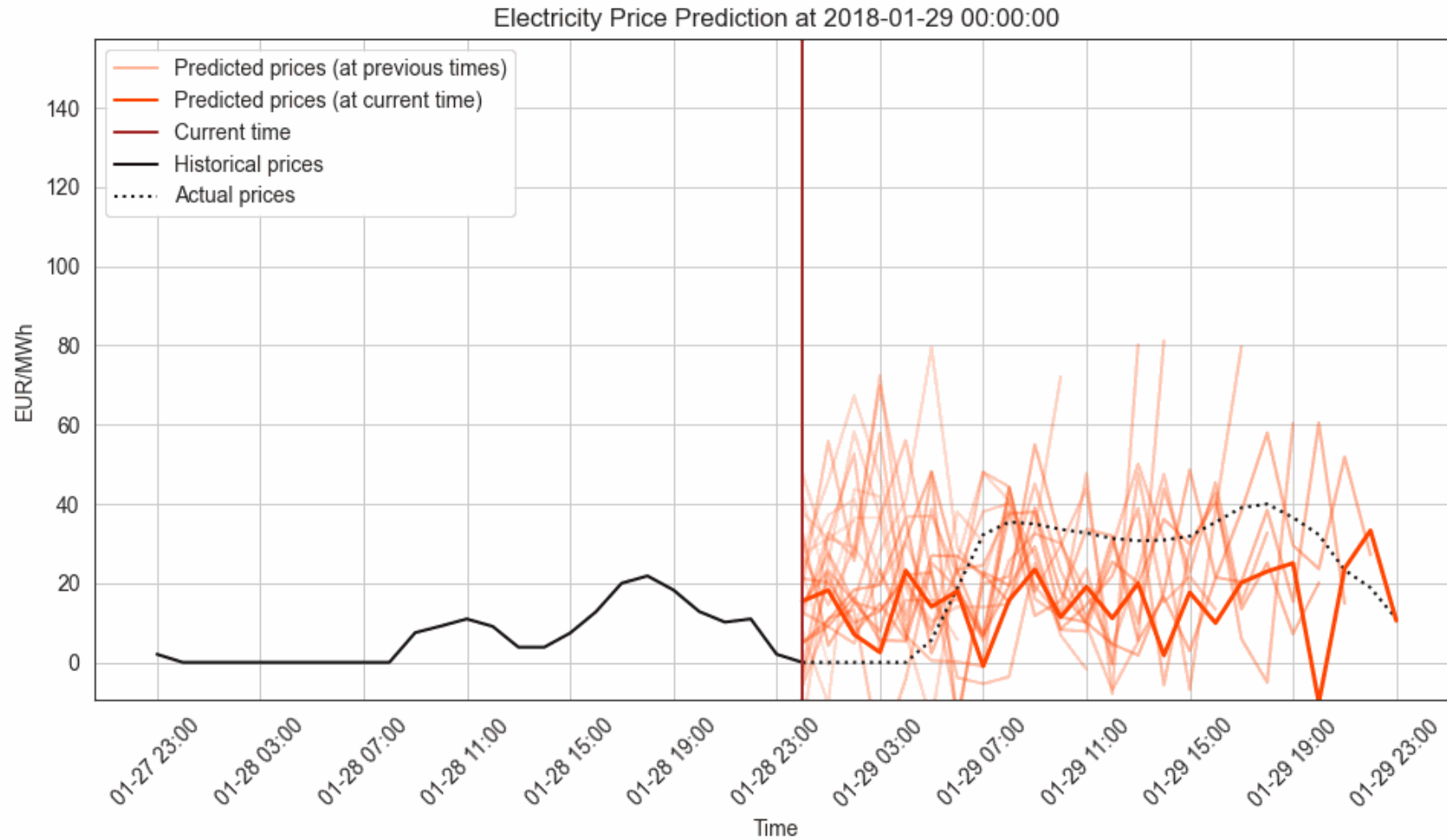
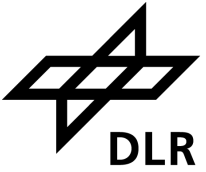
Schimeczek et al. (2023a). [10.21105/joss.05041](https://doi.org/10.21105/joss.05041)

Schimeczek et al. (2023b). [10.21105/joss.05087](https://doi.org/10.21105/joss.05087)

Nitsch et al. (2023a). [10.21105/joss.04958](https://doi.org/10.21105/joss.04958)

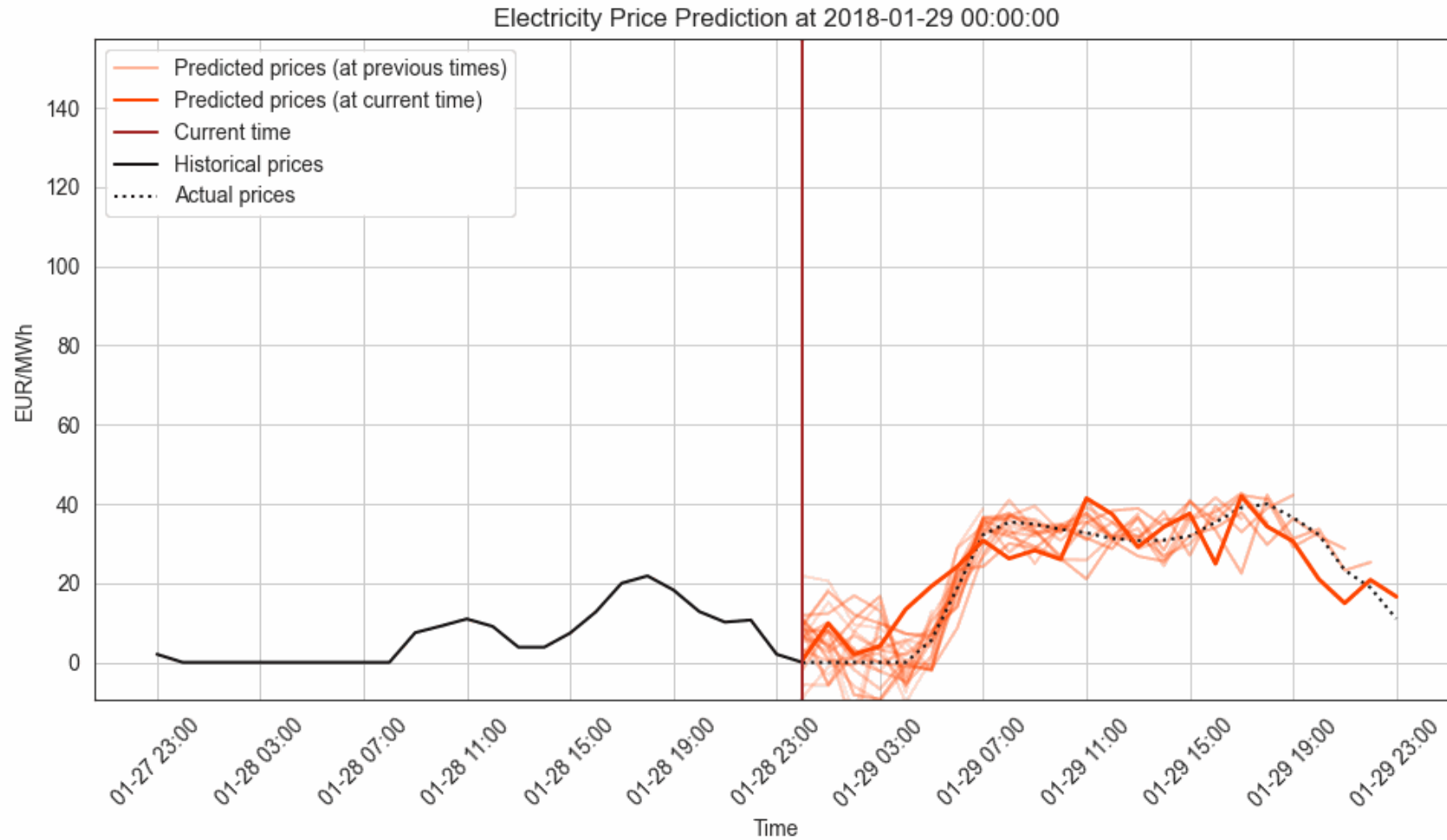
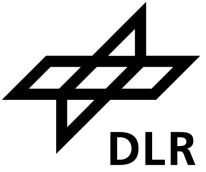
Electricity Price Forecasts

Accuracy – 10 Epochs



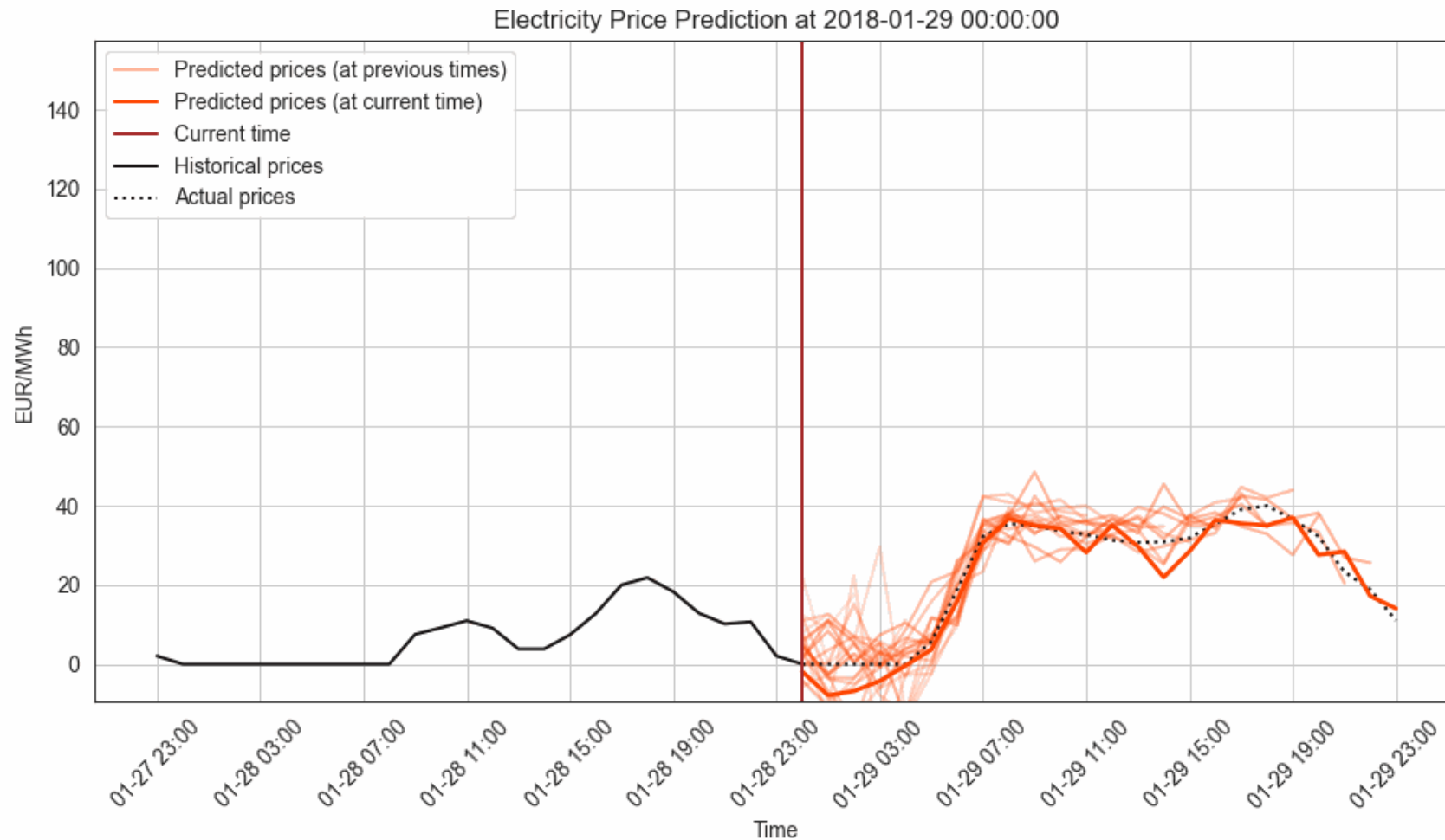
Electricity Price Forecasts

Accuracy – 30 Epochs



Electricity Price Forecasts

Accuracy – 50 Epochs

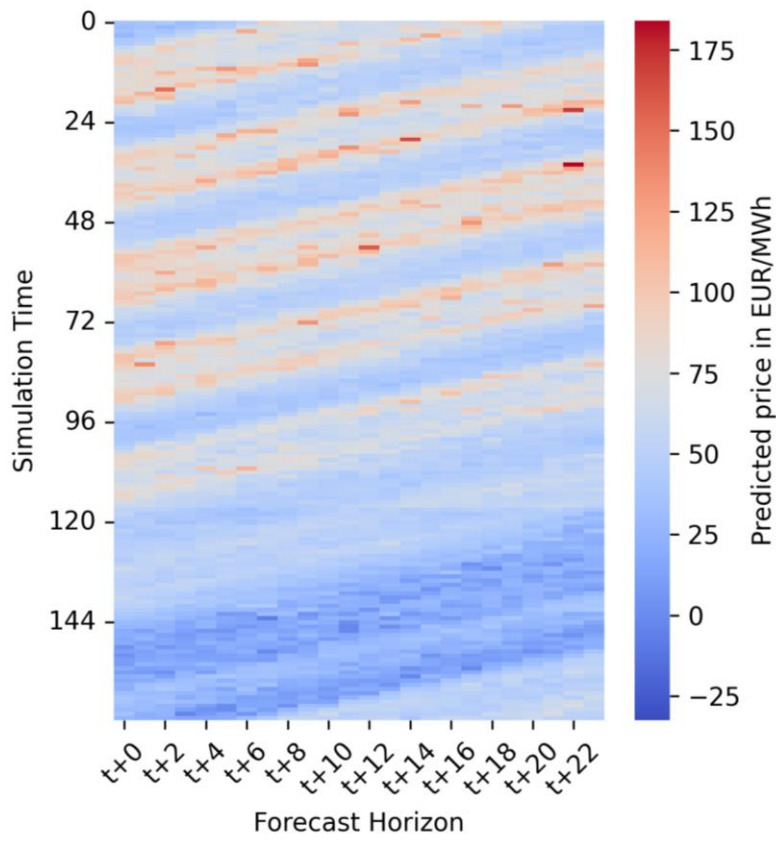


Electricity Price Forecasts

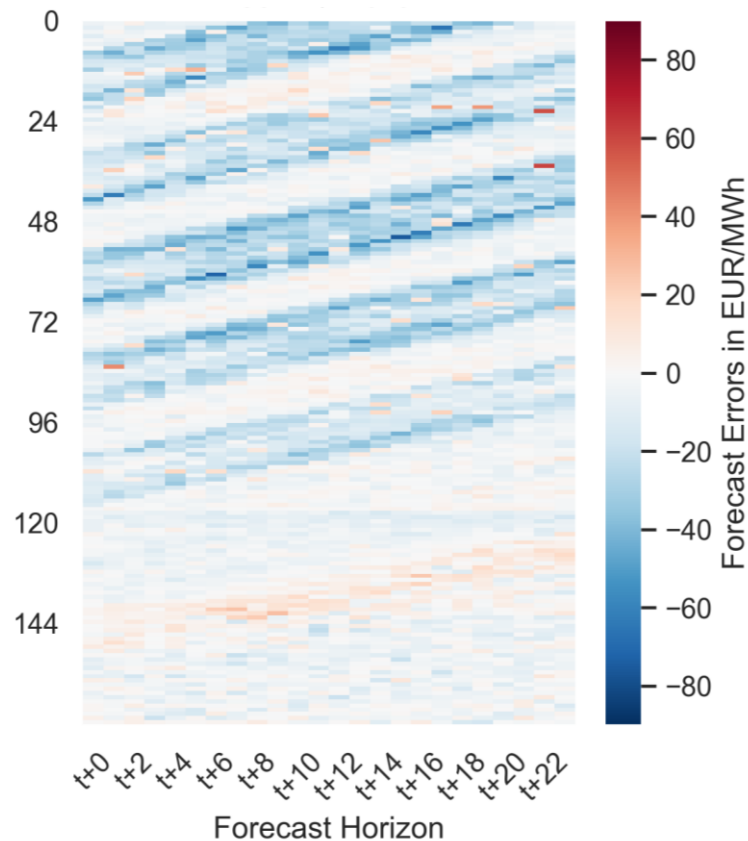
Accuracy – Forecasted Price



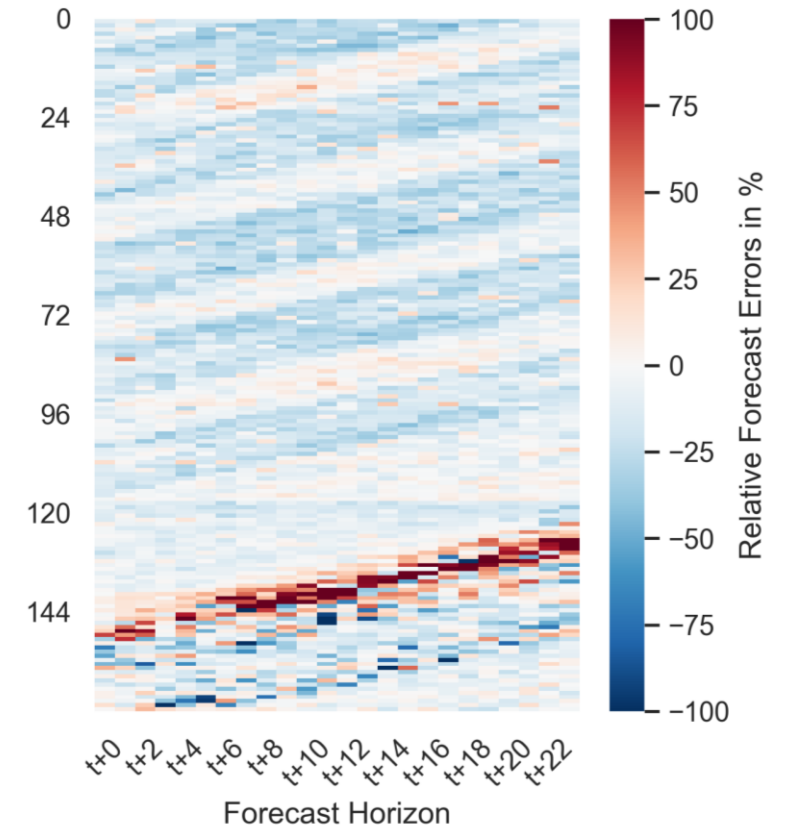
Forecasted Price



Absolute Forecast Error

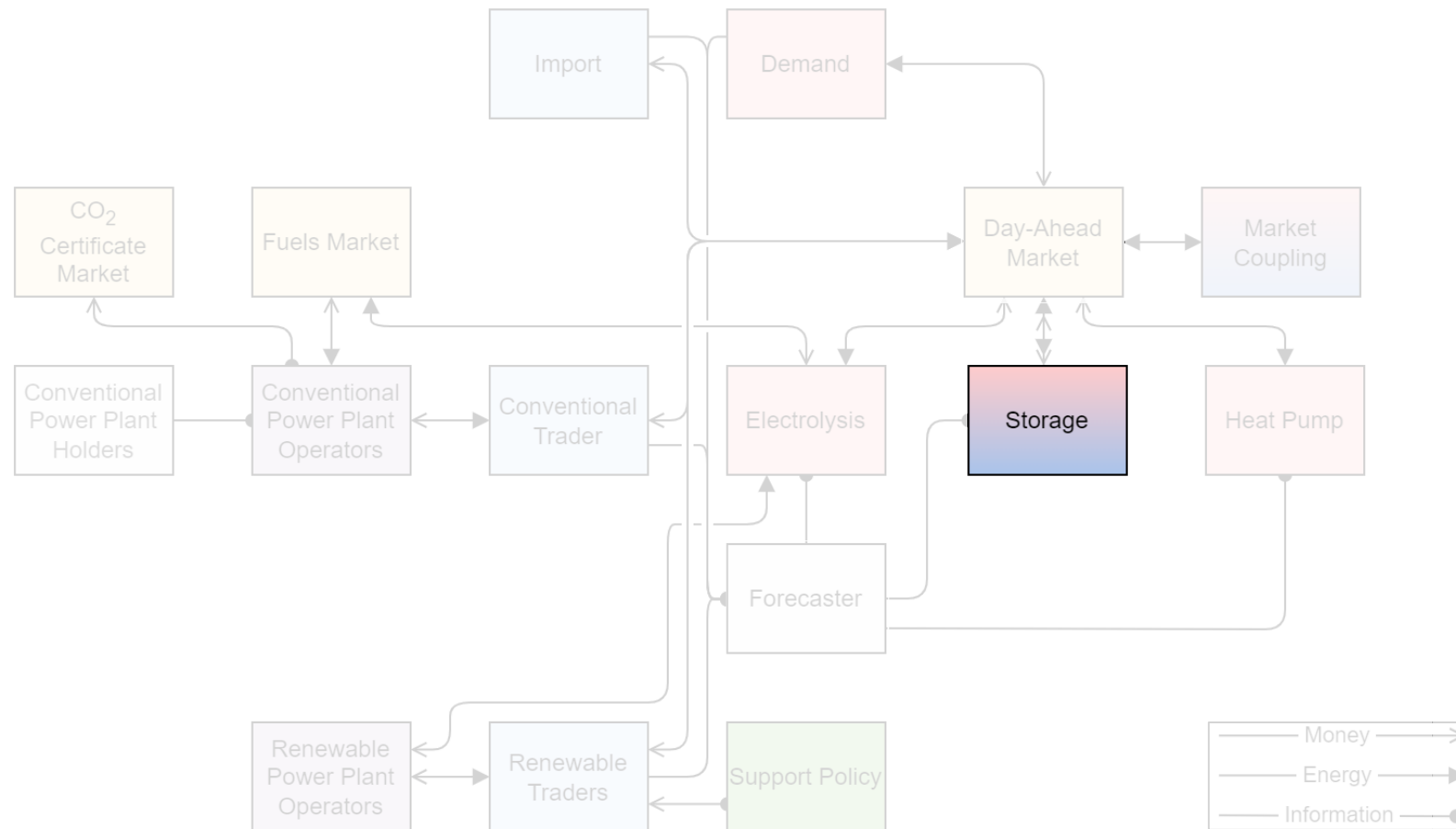
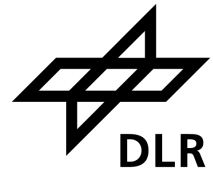


Relative Forecast Error



AMIRIS

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Schimeczek et al. (2023a). [10.21105/joss.05041](https://doi.org/10.21105/joss.05041)

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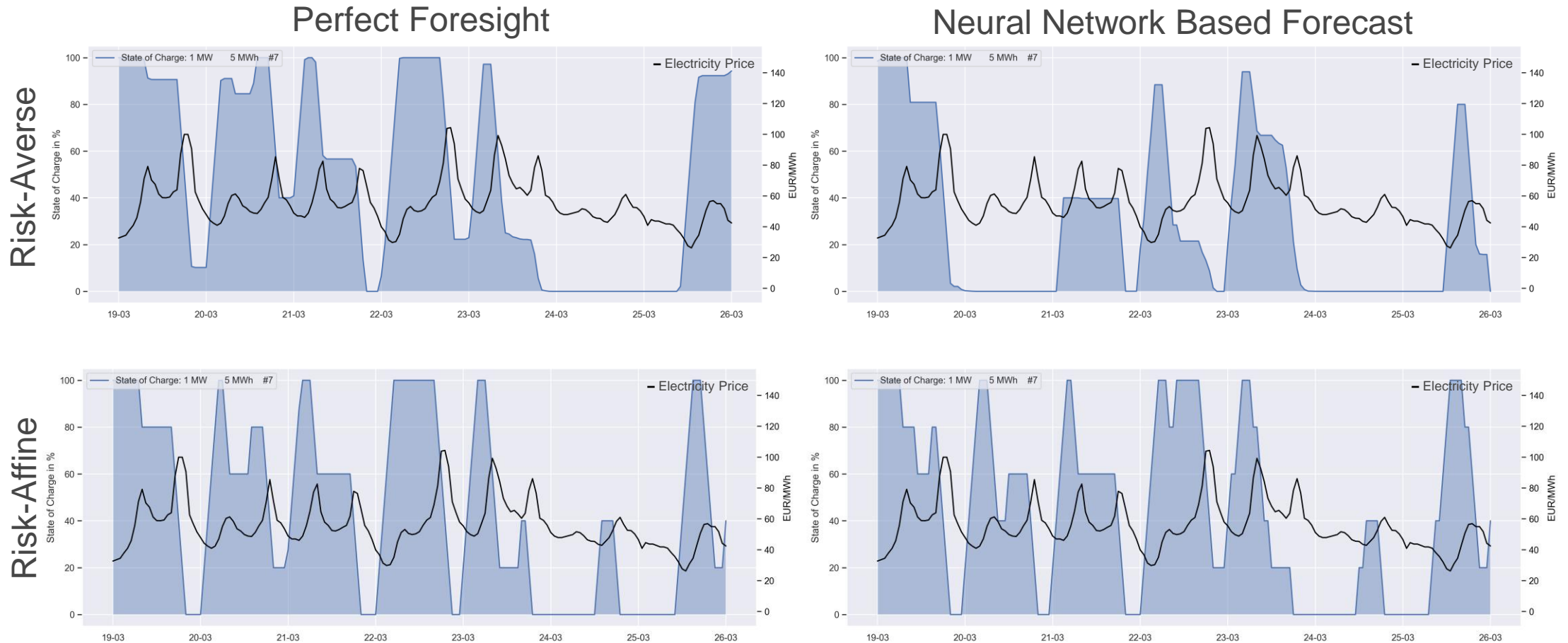
Nitsch et al. (2023a). [10.21105/joss.04958](https://doi.org/10.21105/joss.04958)

State of Charge

Impact of Forecast Type & Storage Strategy



Electricity Price Forecast Type



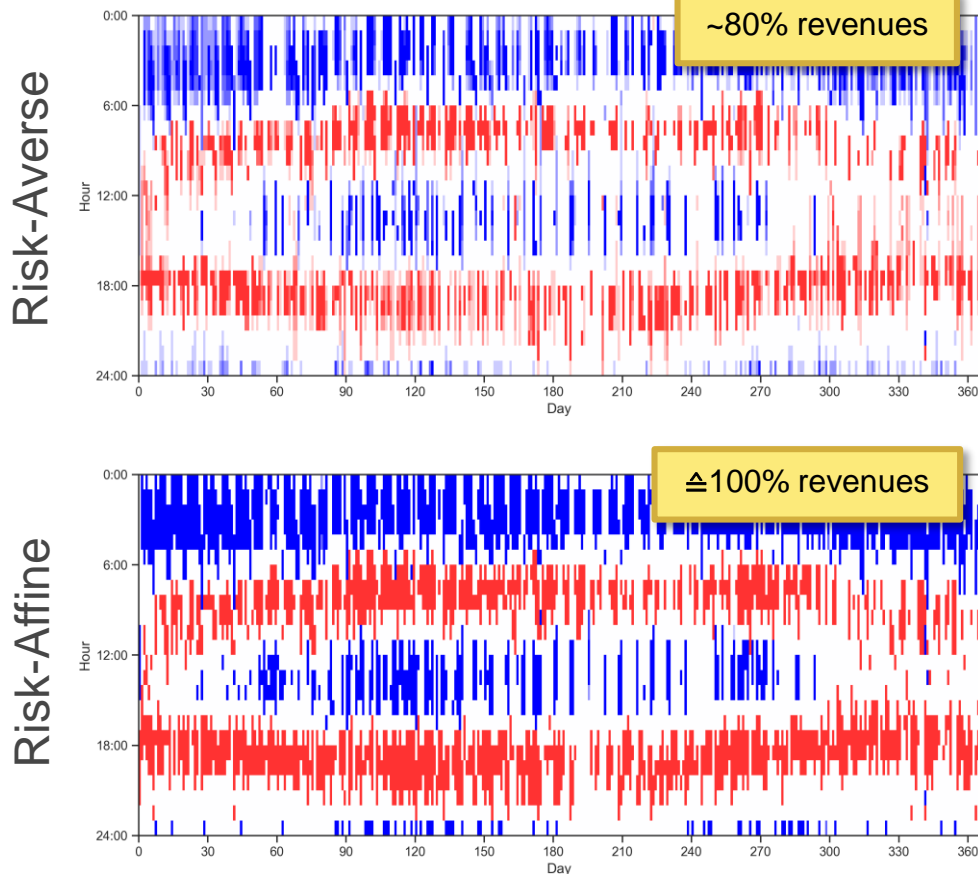
Storage Activity

Impact of Forecast Type & Storage Strategy

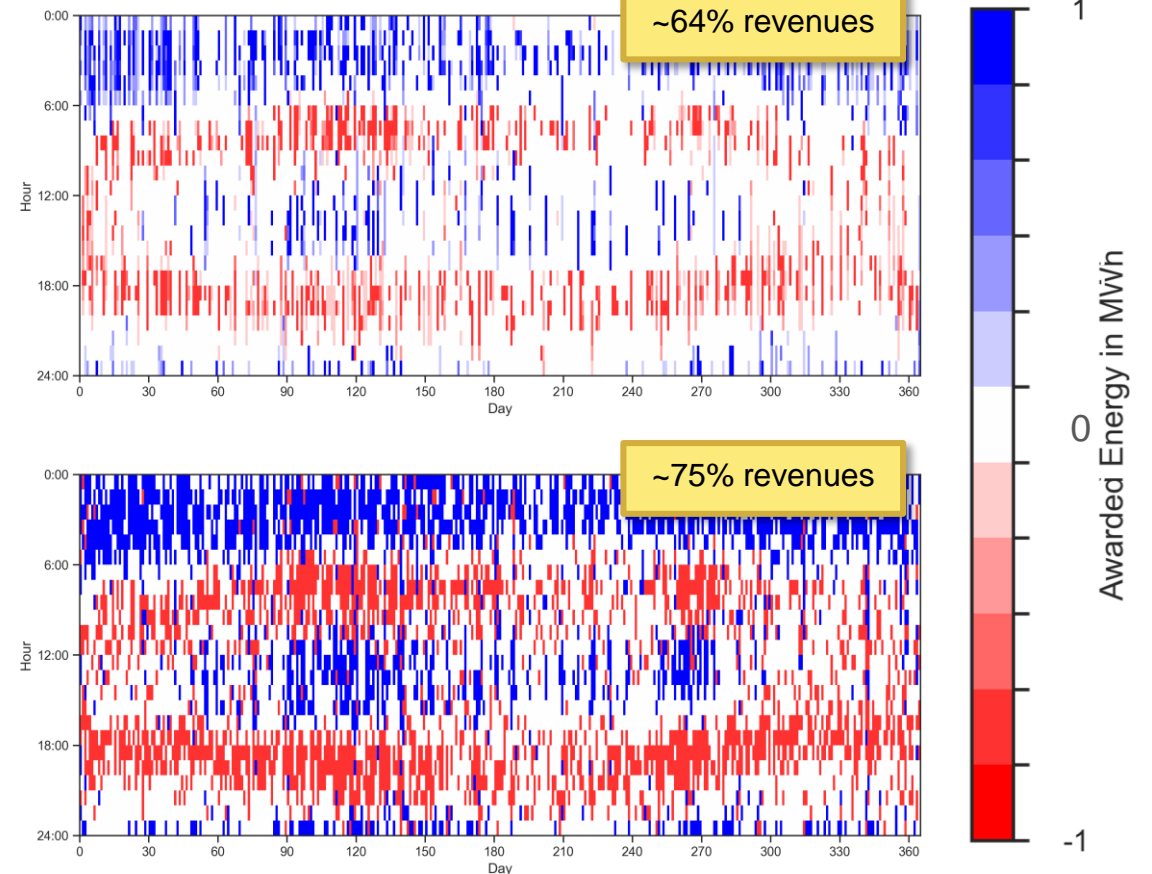


Electricity Price Forecast Type

Perfect Foresight

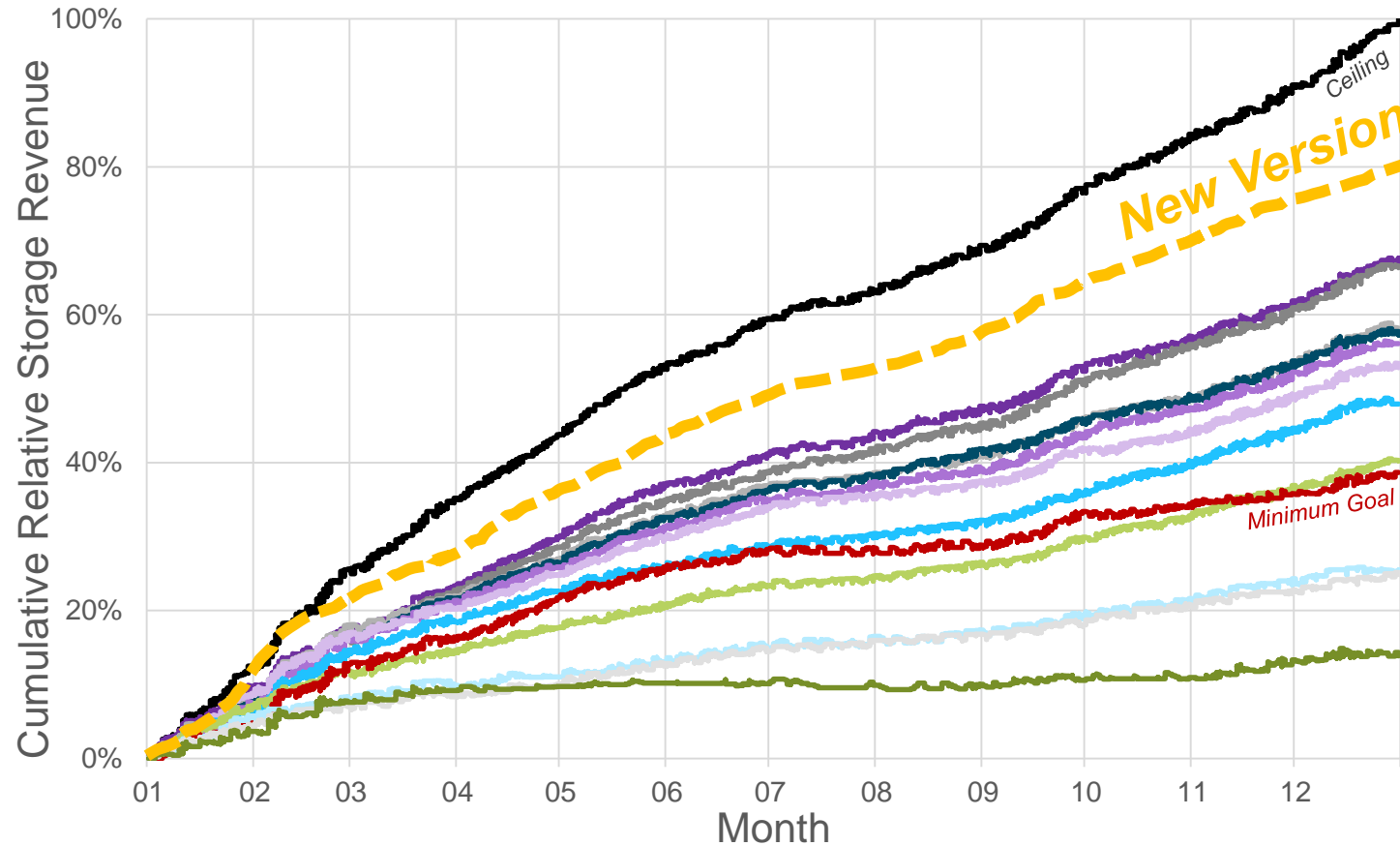


Neural Network Based Forecast



Revenue Impact of the Model Used

The more complex the better?



Model

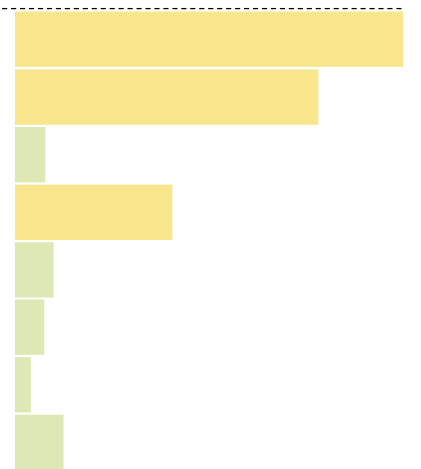
- Perfect Foresight
- TFT 100-Epochs 4-LSTM 32-Hidden
- TFT 100-Epochs 4-LSTM
- TFT 10-Epochs 4-LSTM
- TFT 100-Epochs
- TFT 10-Epochs 4-LSTM 32-Hidden
- TFT 10-Epochs 4-LSTM 8-Hidden
- TFT 10-Epochs
- TFT 10-Epochs 8-LSTM
- Naive TimeShift24
- TFT 1-Epoch
- TFT 1-Epoch 4-LSTM
- TFT 10-Epoch 4-LSTM 48-Lags

TFT Default: 1-LSTM 16-Hidden 24-Lags

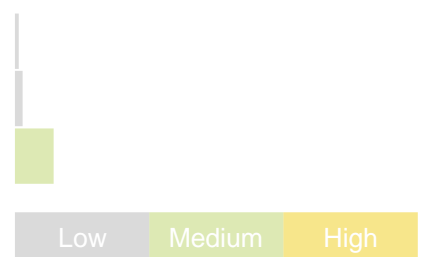
"New version" uses residual load as covariate

Training Effort

n.a.



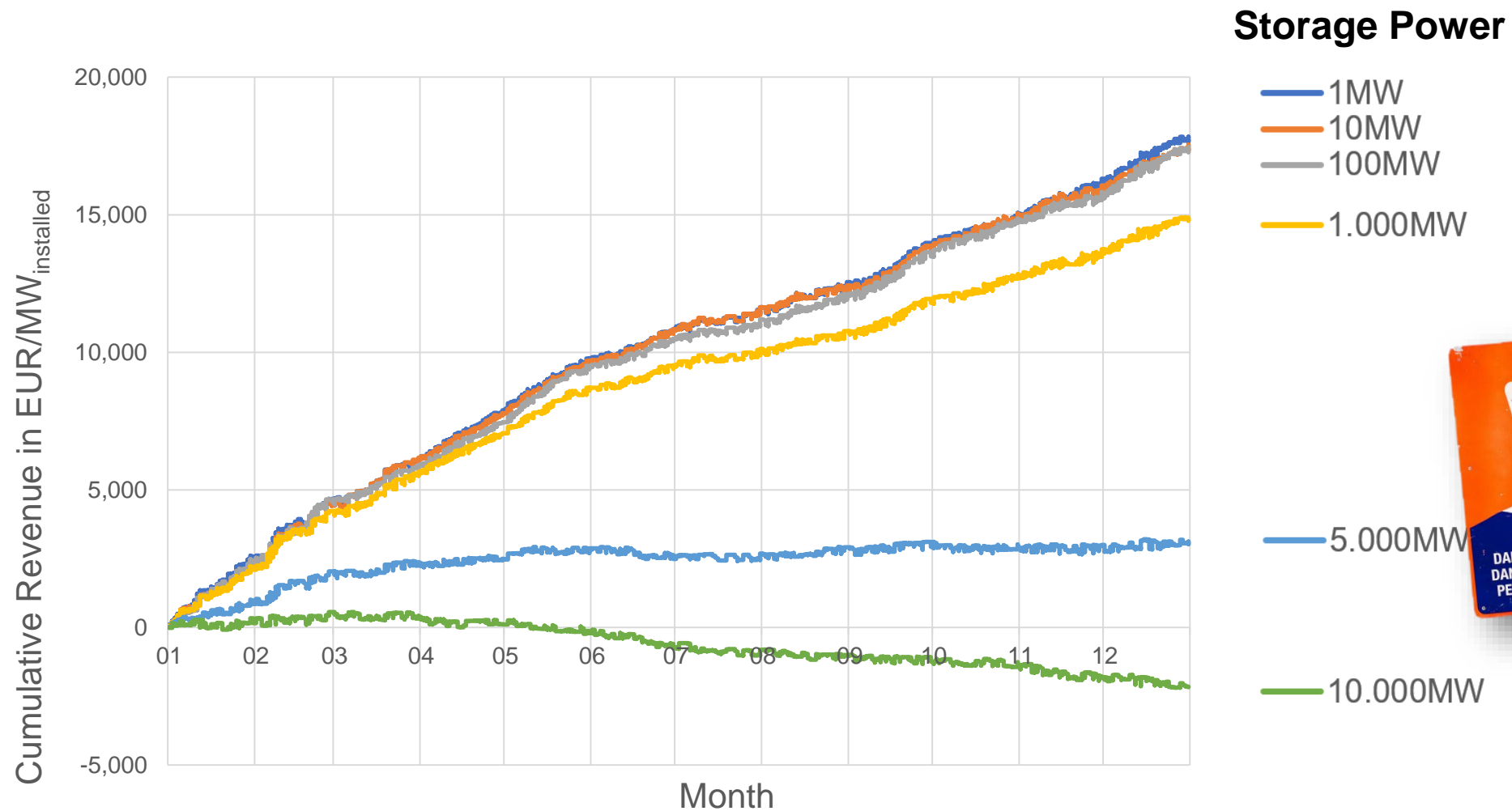
n.a.



Low Medium High

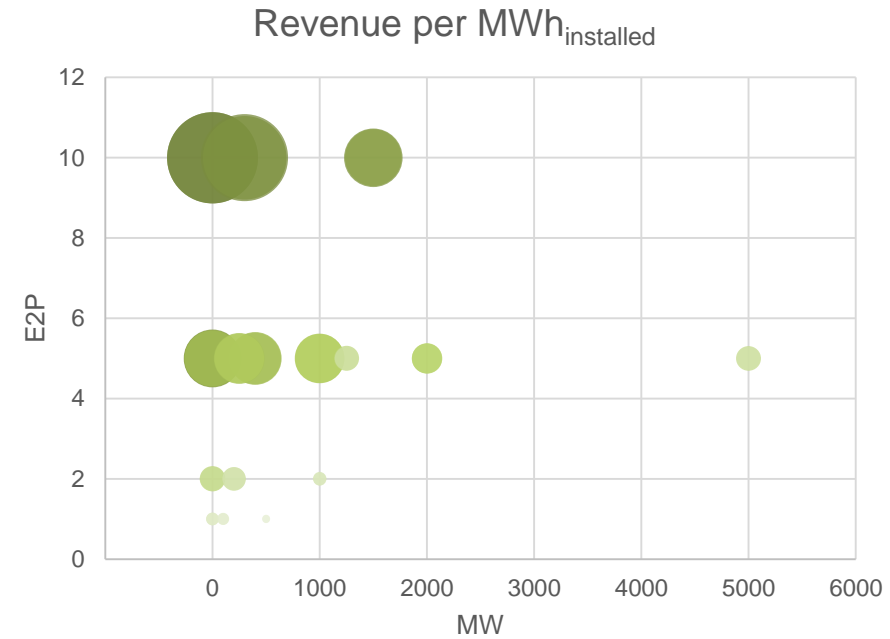
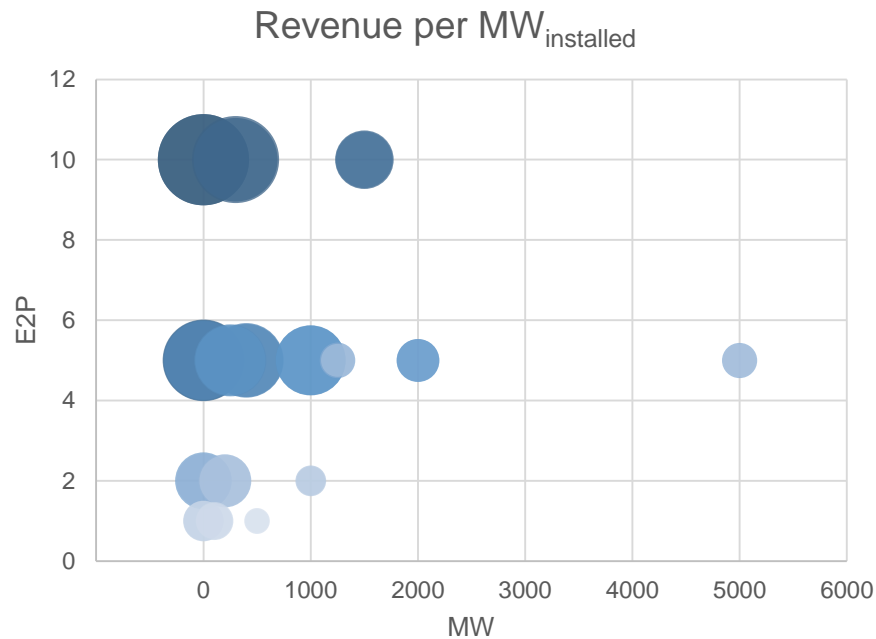
Revenue by Storage Size

Too big to fail for success



Revenue by Storage Specification

Power vs. capacity



Discussion

Novelty and Limitations



Strengths

- Market driven economic evaluation of storage potential
- Fully automated open pipeline
- Accurate and flexible forecasts at runtime
- Fast execution with low overhead

Weaknesses

- No additional revenue streams besides Day-Ahead market (DAM)
 - DAM considered as benchmark/indicator
- Uncertainty in system design
 - Transparent communication of assumptions
 - Open modelling setup and data publication

Conclusions



- Machine learning based electricity price forecasts with **high accuracy**
- **Modular** integration of forecasts in agent-based electricity market simulation
- **Revenue analysis** of storage systems

Outlook

- ARIADNE scenario parameterisation and analysis <https://ariadneprojekt.de/>
- Competition among heterogeneous storage systems
- Publication of the presented tool chain

Acknowledgements: Valentin Bertsch, Kristina Nienhaus on behalf of the Energy Economics Group

Contact: Felix Nitsch, felix.nitsch@dlr.de, German Aerospace Center, Institute of Networked Energy Systems, Energy Systems Analysis

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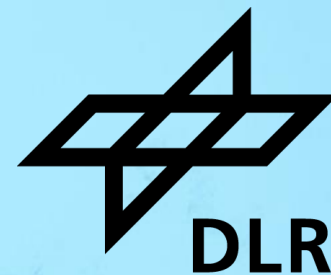


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APPENDIX



AMIRIS

Following FAIR4RS Principles



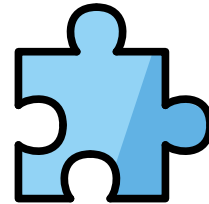
Findable

- [Website](#)
- [DOI](#)
- [Wikipedia](#)
- [COMSES](#)
- [HECI](#)
- [OEP](#)
- [openmod](#)



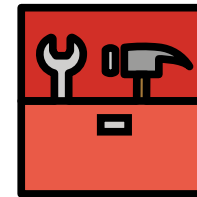
Accessible

- [GitLab](#)
- [PyPI](#)
- [Zenodo](#)



Interoperable

- [API](#)
- [Workflow tools](#)
- [CSV](#)
- [YAML](#)



(Re-)usable

- [Apache 2.0](#)
- [REUSE](#)
- [Wiki](#)
- [Javadoc](#)
- [Win/Mac/Linux](#)
- [Scalable \(H\)PC](#)

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Key Indicators



Users

- 25 confirmed external user
- 9 external contributions



PhD candidates

- 7 internal
- 6 external



Visibility

- 25k views on Wikipedia
- 16k views on openmod

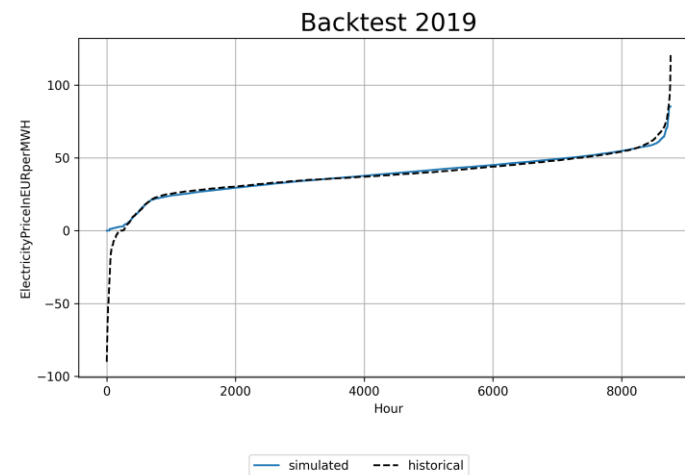
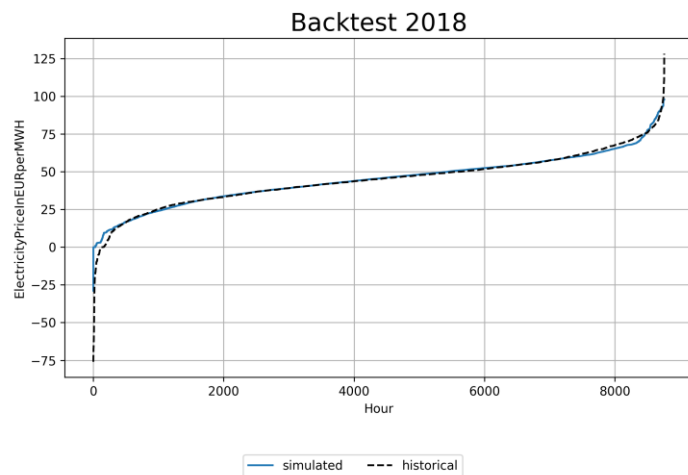
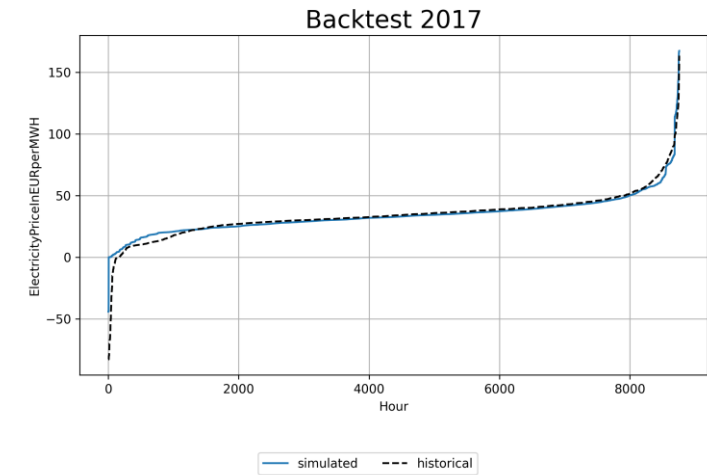
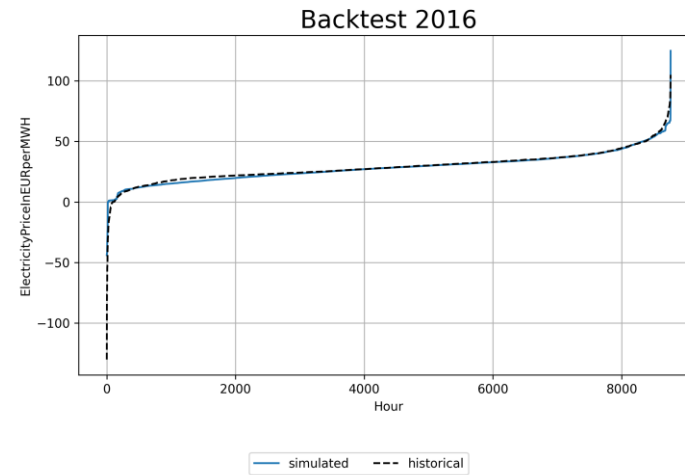
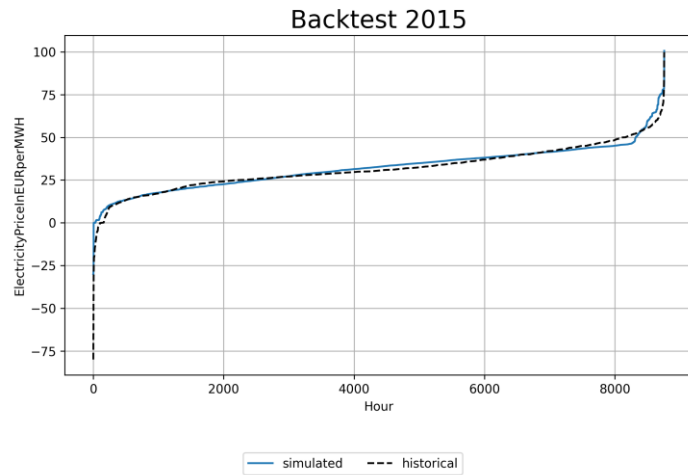


Software

- 35 releases
- 27k downloads

AMIRIS Backtesting 2015 – 2019

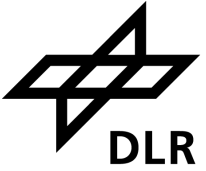
Price duration curves



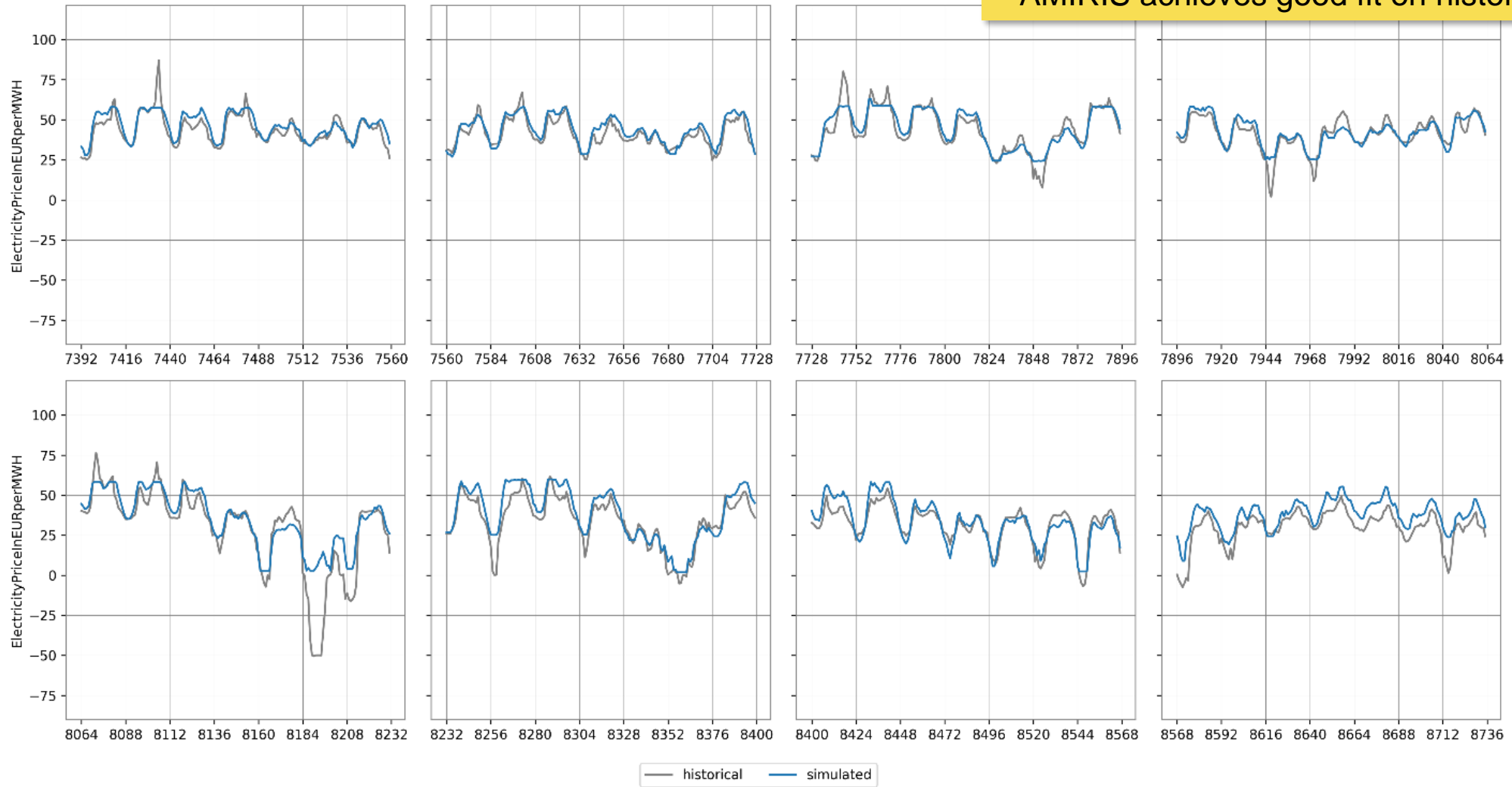
- Model parameterization using open data only
- Benchmarking and convenient scenario setup
- Data available in open [AMIRIS example repository](https://zenodo.org/record/7789049)

AMIRIS Backtesting 2015 – 2019

Exemplary weeks in 2019




AMIRIS achieves good fit on historical data



Imprint



Topic	ML-Based Price Forecasts in the open electricity market model AMIRIS
Date	2025-02
Author	Felix Nitsch
Institute	Institute of Networked Energy Systems
Credits	DLR (CC BY-NC-ND 3.0)  except stated otherwise

Acknowledgements: Valentin Bertsch, Kristina Nienhaus on behalf of the Energy Economics Group

Felix Nitsch, Institute of Networked Energy Systems, 2025-02

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