

# Optimized Operation of Active Distribution Grids with a Control System based on Graph Neural Networks

Manuela Linke<sup>a</sup>, Marcel Arpogaus<sup>a</sup>, Lea Anat Brass<sup>a</sup>, Julian Maier<sup>a</sup>, Christof Wittwer<sup>b</sup>, Gunnar Schubert<sup>a</sup>

<sup>a</sup>University of Applied Sciences Konstanz HTWG, Konstanz, Germany

<sup>b</sup>Fraunhofer Institut for Solar Energy Systems ISE, Freiburg, Germany

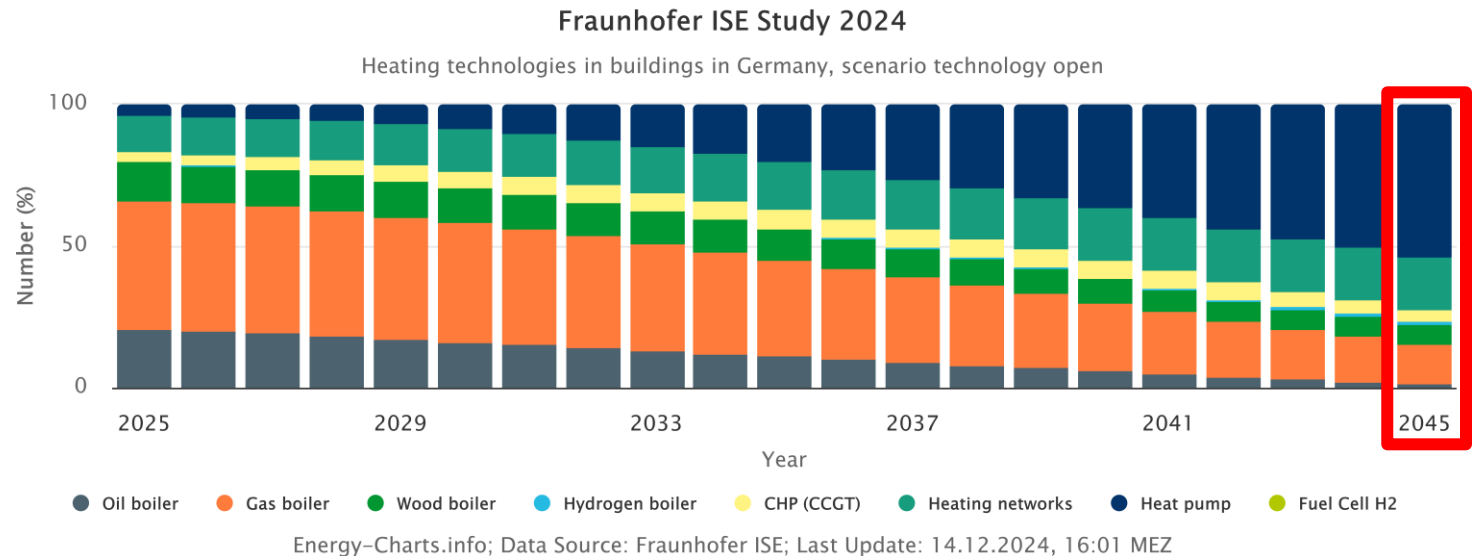
ABM4Energy, 31.03.2025

# Motivation

## Challenges of the energy transition in Germany

### Electrification of the consumer sector by 2045

- Share of heat pumps > 50%
- Up to 35 million electric vehicles



- Increasing amount of volatile generators in the grid
- Distribution grids are the least monitored and controlled parts of the grid
- Conventional methods (e.g. state estimation) are not suitable

Image source: [www.energy-charts.info](http://www.energy-charts.info) - last update: 14.12.2024, 16:01 MEZ

Data source: Thelen, C. et al. „Wege zu einem klimaneutralen Energiesystem: Bundesländer im Transformationsprozess“, 2024, Fraunhofer ISE

# Grid operation management

## Goal

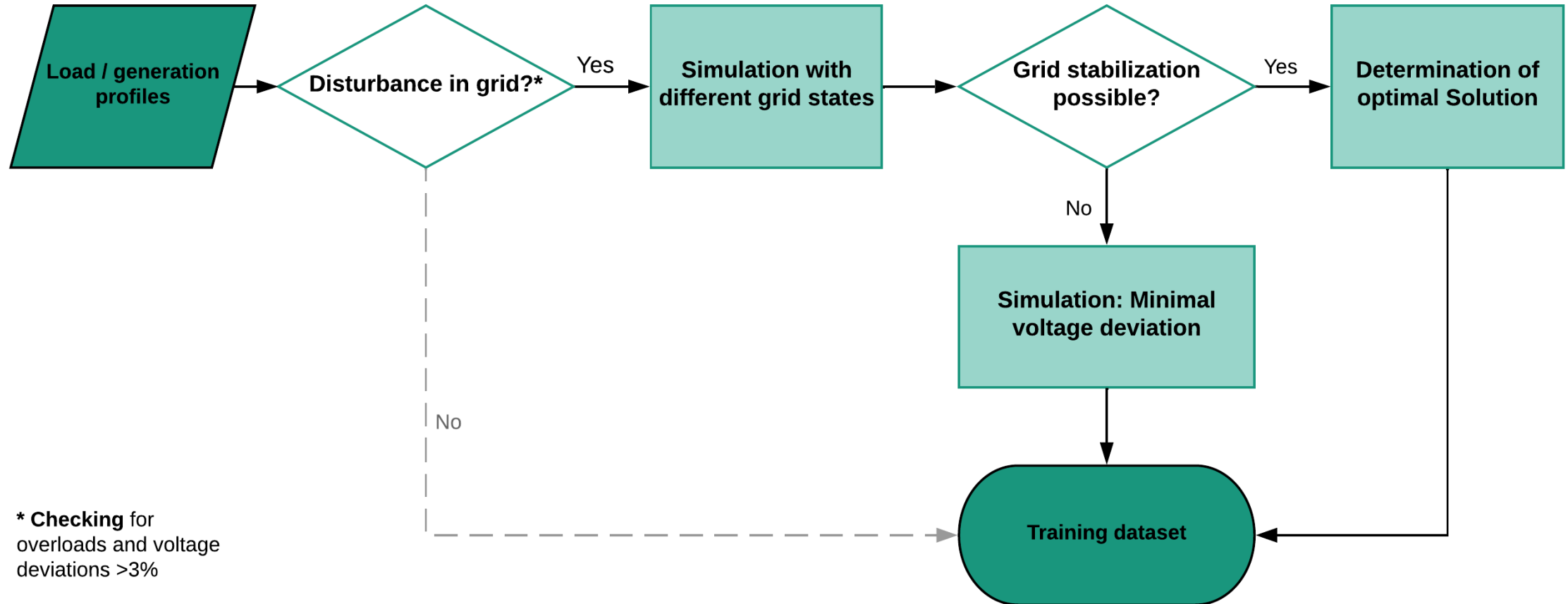
- Intelligent use of the existing grid infrastructure
- Reduction on grid expansion
- Avoidance of supply bottlenecks

## Solution measures

- **Transformer tap changers**
- Congestion measures of controllable loads and generators
- Remote controlled switches

**Problem:** Data availability and quality!

# Training data generation

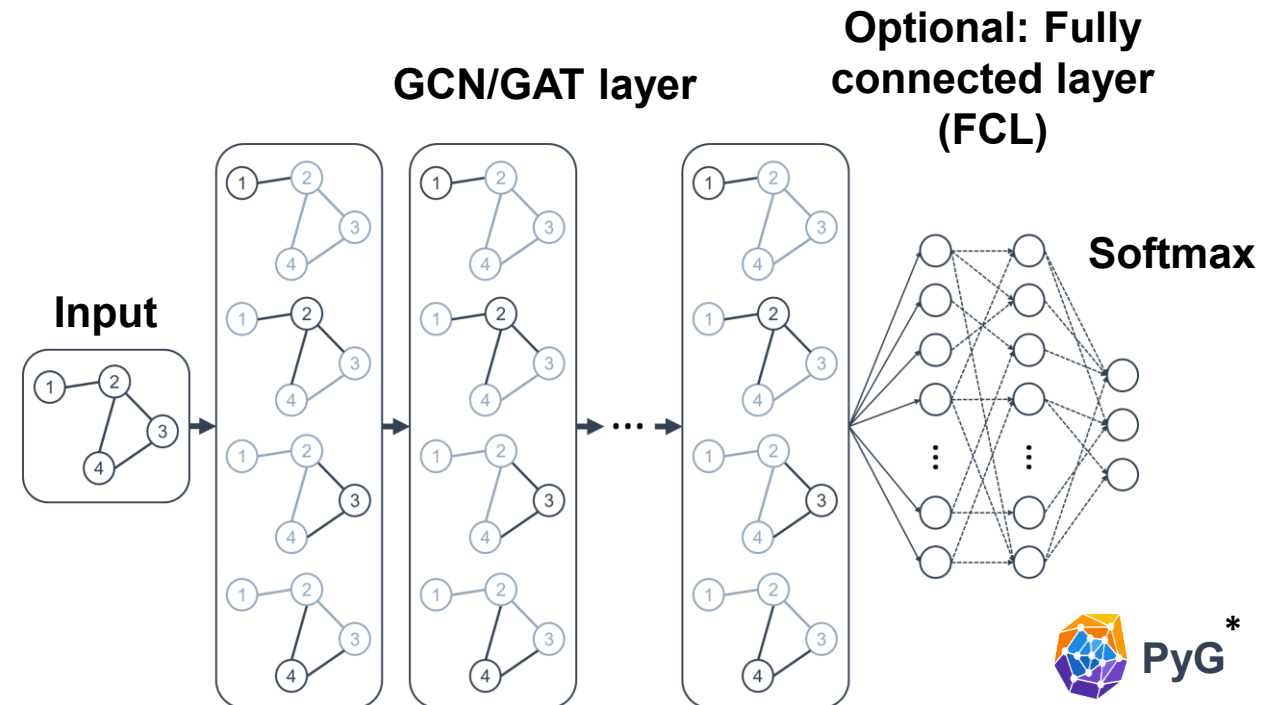


\* Checking for overloads and voltage deviations >3%

# Graph Neural Network

Graph Convolutional Network (GCN) / Graph Attention Network (GAT)

- Topology of the network is used for training (neighborhood relationships)
- Independent of the size of the input and permutation invariant
- Generalizability: Trained GNN can also be applied to new topologies

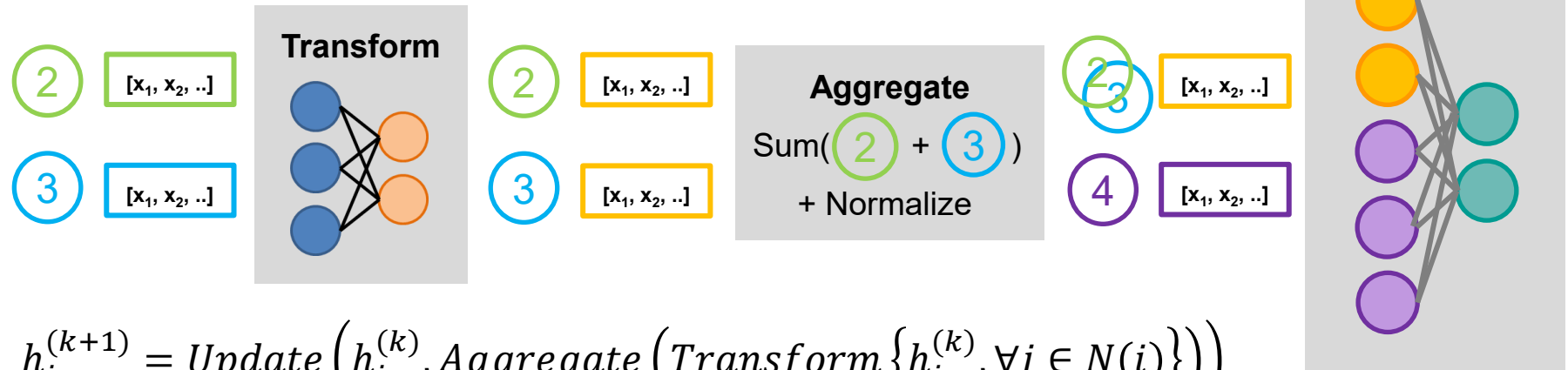
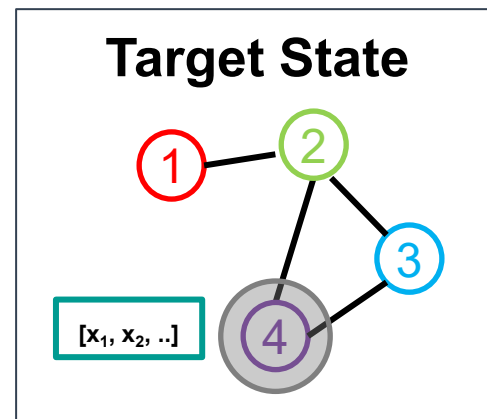
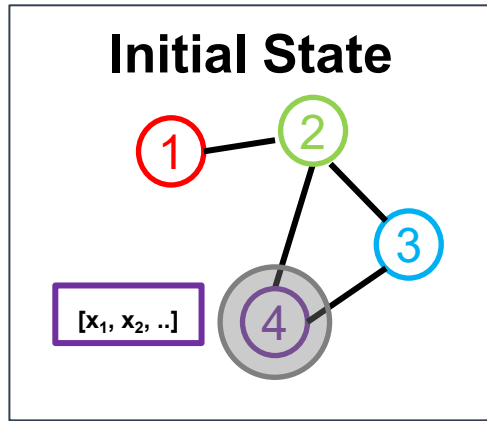


M. Linke et al., „Distribution Grid Management with Graph Neural Networks”, <https://www.zee-uni-freiburg.de/wp-content/uploads/2023/11/Poster-12-Distribution-grid-management-with-graph-neural-networks.pdf>

\* M. Fey, J. E. Lenssen, Fast Graph Representation Learning with PyTorch Geometric, 2019, ICLR 2019 (RLGM Workshop), <https://doi.org/10.48550/arXiv.1903.02428>

# Graph Neural Networks

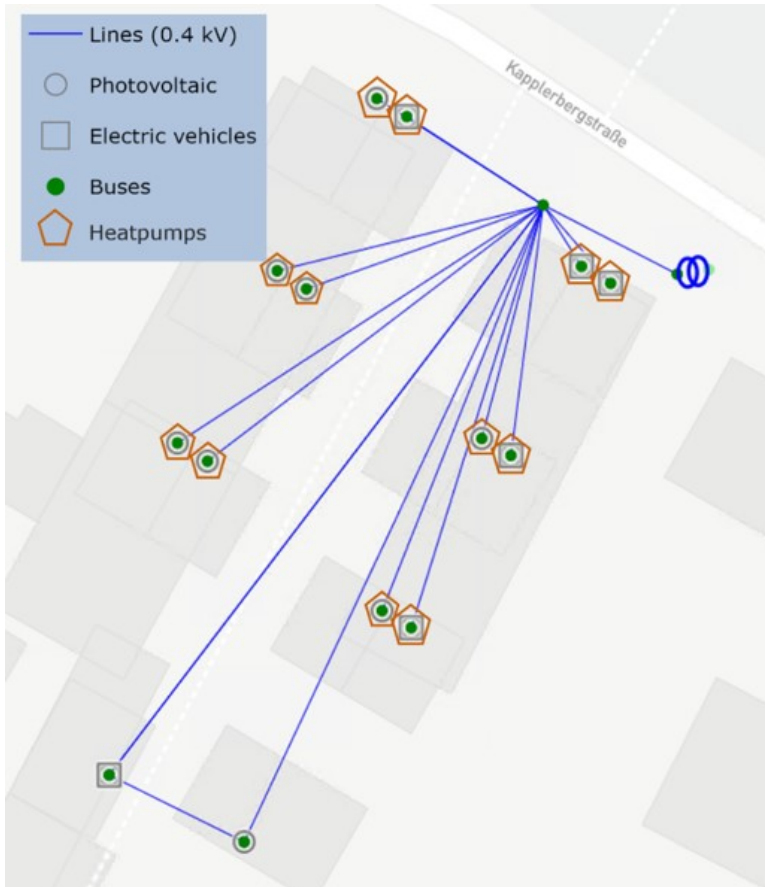
## Message Passing Mechanism



$$h_i^{(k+1)} = \text{Update} \left( h_i^{(k)}, \text{Aggregate} \left( \text{Transform} \{ h_j^{(k)}, \forall j \in N(i) \} \right) \right)$$

# Test

## LV grid with 19 nodes



## Grid parameters

Node features	Loads [MW]																
	Generations [MW]																
Edge features	Line length [km]																
	Primary line constants $(x, r, g, b)$																
	Tap ratio of transformer station																
Edge index	Adjacency matrix																
	<table border="1"><tr><td>0</td><td>1</td><td>...</td><td>1</td></tr><tr><td>1</td><td>0</td><td>...</td><td>0</td></tr><tr><td>...</td><td>...</td><td>...</td><td>...</td></tr><tr><td>1</td><td>0</td><td>...</td><td>0</td></tr></table>	0	1	...	1	1	0	...	0	...	...	...	...	1	0	...	0
	0	1	...	1													
	1	0	...	0													
	...	...	...	...													
1	0	...	0														

# Results

Graph Convolutional Network (GCN) / Graph Attention Network (GAT)

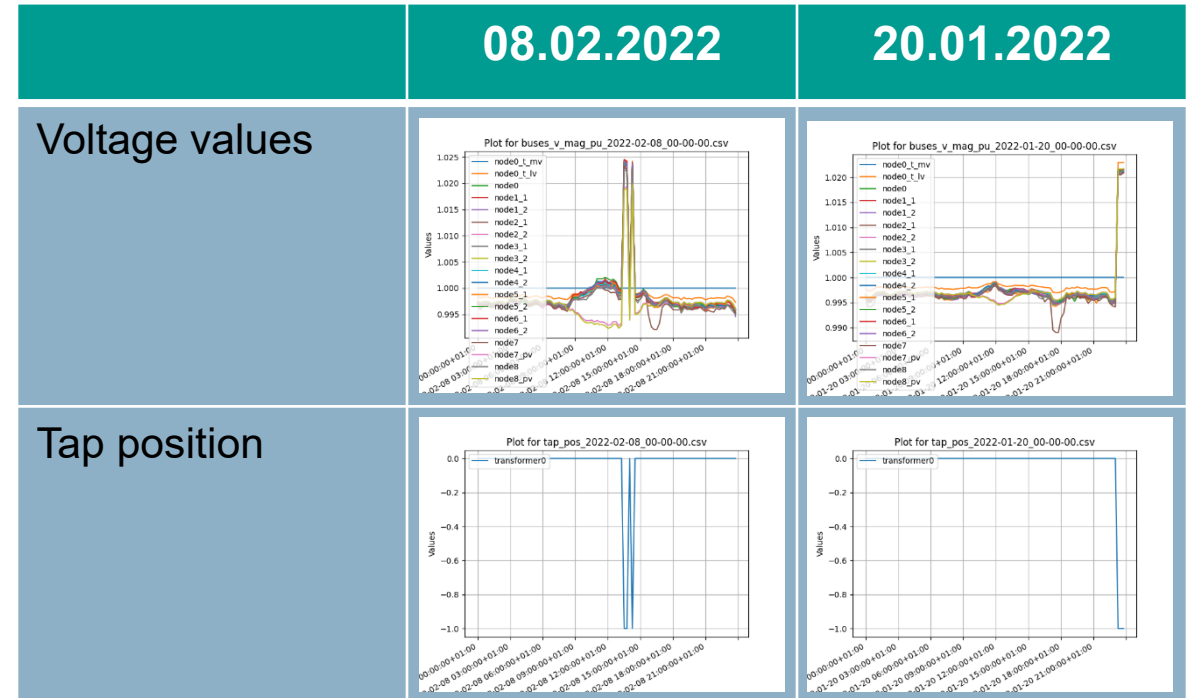
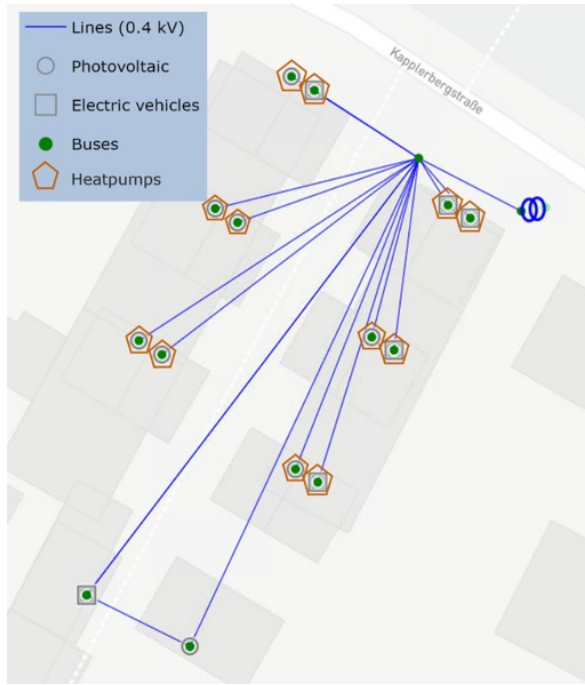
Model	Training Loss	Validation Loss	Training Accuracy	Validation Accuracy	Test Accuracy
GCN	0.191	0.152	0.945	0.944	0.922
GAT	<b>0.099</b>	<b>0.079</b>	<b>0.964</b>	<b>0.976</b>	<b>0.944</b>
GCN-FCL	0.238	0.164	0.949	0.933	0.930
GAT-FCL	0.097	0.108	0.968	0.981	0.936

FCL: Fully Connected Layer



# Tested at the Digital Grid Lab @ Fraunhofer ISE\*

- Electricity grid: Digital twin of the Allensbach electricity grid
- Time series of 3 months (January to March 2022)
- 4 voltage band violations detected



\* <http://www.digital-grid-lab.de/>

# Conclusion & Outlook

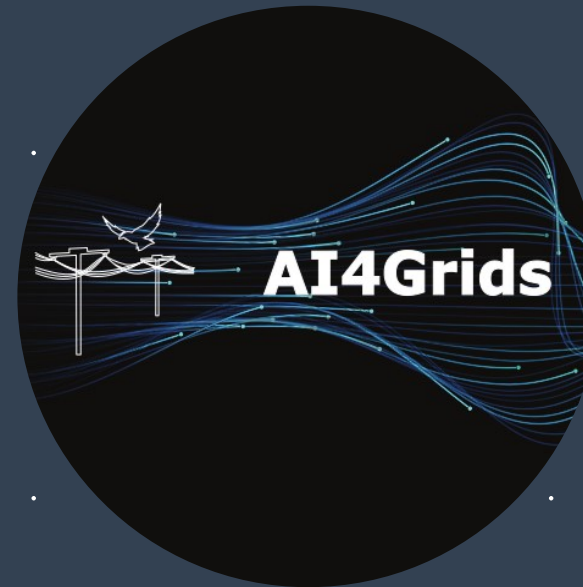
- GNNs are well-suited for controlling a distribution grid and enable optimal use of existing grid infrastructure
- Data distribution of the classes affects accuracy of the prediction
- Tested in lab environment

## Outlook

- Congestion measures of controllable loads and generators
- Explainability to meet the requirements of the EU AI Act\*
- Scalability: Combination with agent-based system to realize in large grid areas
- Application to the dynamic and complex energy-relevant processes for process optimization in the energy-intensive industry

\*European Parliament and Council of the European Union, Regulation (eu) 2024/1689 of the european parliament and of the council of 13 June 2024 laying down harmonised rules on artificial intelligence and amending regulations (ec) no 300/2008, (eu) no 167/2013, (eu) no 168/2013, (eu) 2018/858, (eu) 2018/1139 and (eu) 2019/2144 and directives 2014/90/eu, (eu) 2016/797 and (eu) 2020/1828 (artificial intelligence act) (2024). URL <http://data.europa.eu/eli/reg/2024/1689/oj>

H T  
W  
G



Hochschule Konstanz  
University of Applied Sciences

**Thank you for your attention!**

manuela.linke@htwg-konstanz.de



Supported by:



Federal Ministry  
for the Environment, Nature Conservation,  
Nuclear Safety and Consumer Protection

based on a decision of  
the German Bundestag