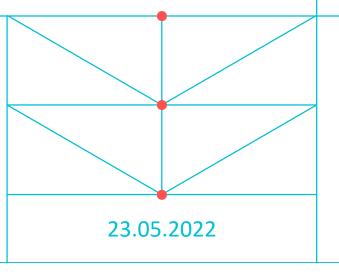
# Predicting Regional Soil Moisture Dynamics Using Machine Learning Techniques and A Dense Observational Network



# TUHH Technische Universität Hamburg





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## Introduction & Objectives



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- Soil moisture has a direct impact on ecosystem functioning, vegetation and crop production and plays a crucial role in all aspects of landatmosphere interactions.
- The highly localized and complex nature of soil moisture present a major challenge to its accurate estimation.
- Notwithstanding recent advances in satellite-based monitoring, the temporal and spatial resolution and shallow observation impede their application to mechanistic modelling and to highly resolved applications.

## **Objectives**

 We aim to develop a predictive tool capable of describing the relationship between soil moisture and a wide range of climatic and soil related parameters using AI-based approaches.

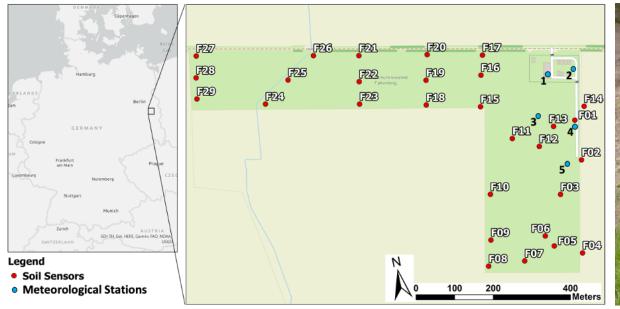


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#### Field Measurement and Data Collection



- We conducted a detailed observational campaign covering about 20,000 m<sup>2</sup> in Falkenberg in Germany as a part of FESSTVal campaign (https://fesstval.de/) using several sensors measuring a wide range of soil and climatic parameters
- Soil sensors recorded soil moisture, temperature and electrical conductivity with temporal resolution of 10 minutes.
- In five meteorological stations, various climatic parameters were recorded including near surface and air temperature<sup>1</sup>, relative humidity<sup>5</sup>, wind speed<sup>3</sup>, precipitation<sup>2</sup>, pressure<sup>1</sup>, turbulent fluxes<sup>4</sup>, solar and terrestrial waves<sup>4</sup>.





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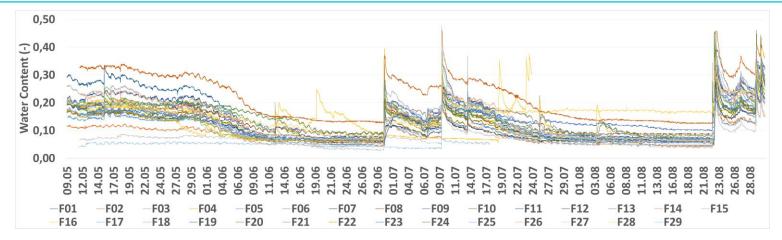
## Typical measured soil data

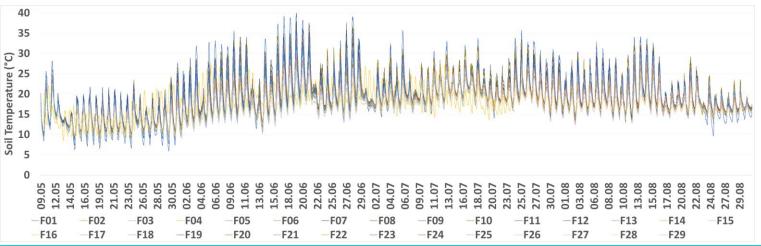


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Typical examples of measured soil water content and temperature over time







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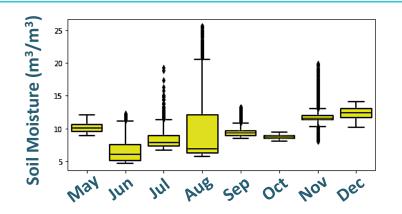
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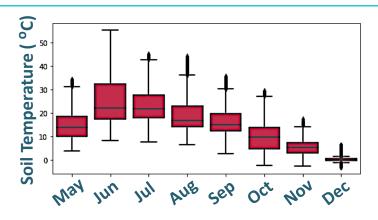
## **Data Analysis**



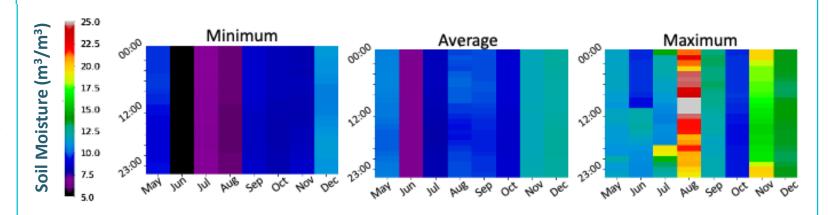
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 Soil moisture and temperature variations during 8 months of year 2021 were analyzed using Python statistical data visualization library, seaborn.





Boxplots and heat maps
 are used to illustrate the
 statistical distribution of
 measured soil moisture &
 temperature and the variations
 of recorded soil moisture at
 different times of the day.
 (averaged over each month)



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## Model Development



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• We used Pearson's coefficient to quantify relationships between the variables with +1 and -1 representing a positive and negative correlation respectively, and 0 representing no relationship

 Long short-term memory (LSTM) model was used in our analysis to capture the complex relationship between the soil moisture and predictors. Based on the computed Pearson's coefficients, 9 variables were selected to train models relating soil moisture (target value) to predictors.

> Model Optimization

Data **Analysis** 

Result

Evaluation

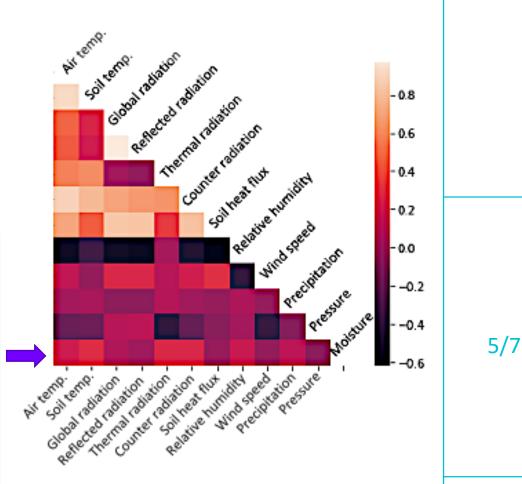
Data

**Processing** 

Model

Construction



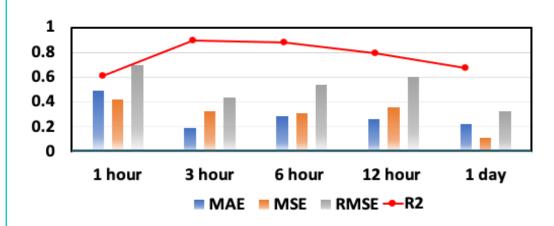


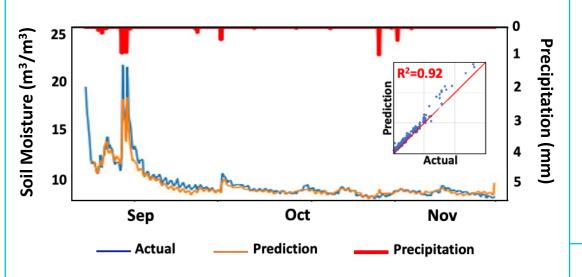
#### **Results & Discussion**



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- Constructed LSTM model was optimized in terms of structure and computational complexity. Feed data density was a key element influencing computational complexity.
- Robust predictions could be obtained when data measured during one day with a time resolution of 3 hours were used to predict the soil moisture one step ahead.
- Various performance indicators were used to evaluate the performance of trained models. The effect of data density was evaluated via four indicators (Mean Absolute Error, Mean Squared Error, RMSE and R<sup>2</sup>). Data density of 3 hours (with R<sup>2</sup> over 90%) was used for the subsequent model training and predictions.
- A typical example of predicted soil moisture dynamics (for one sensor) evaluated against the measured values is illustrated.





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#### Conclusions and Future work



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- All approaches offer a great potential to predict soil parameters and responses
- Predicting the measured soil moisture by all sensors.
- Developing a general AI-based framework to predict spatial and temporal variations of soil moisture under different climatic conditions.

## Thank you!

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