

Unlocking the Potential of ML for Earth and Environment Researchers



Frauke Albrecht, Caroline Arnold, Danu Caus,
Harsh Grover, Andrey Vlasenko, Tobias Weigel

AI Consultants Earth & Environment @DKRZ

Helmholtz AI

Artificial Intelligence Cooperation Unit



Mission

Bring applied AI / ML techniques to your research questions and datasets

6

LOCAL UNITS

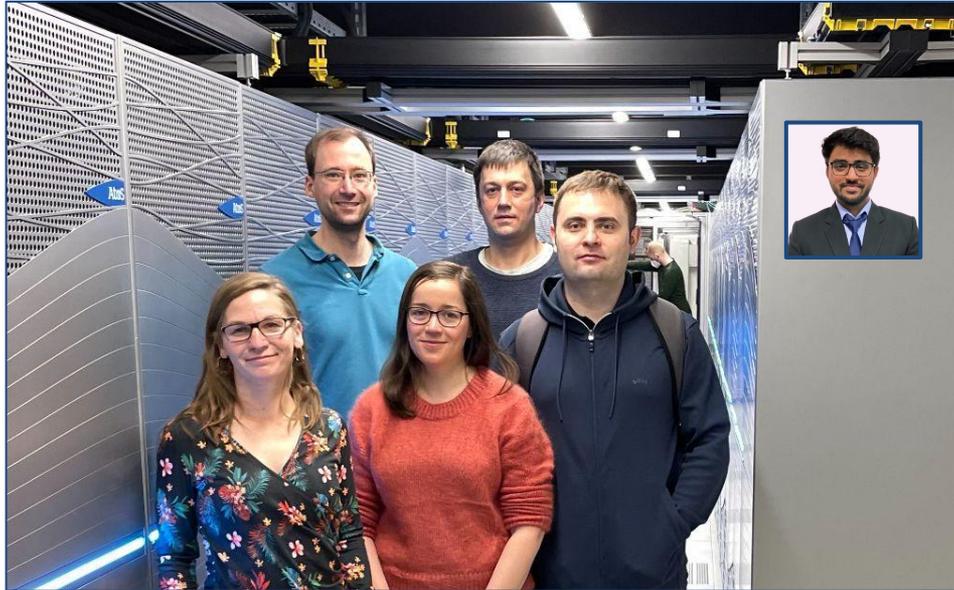
Earth @ HZG
Energy @ KIT
Matter @ HZDR
Space @ DLR
Health @ HMGU
Info @ FZJ

Each Unit

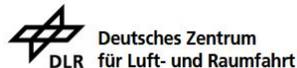
- Young Investigator Group
- AI Consultants



AI Consultants for Earth & Environment



Vouchers from all
Earth & Environment
Helmholtz Centers



Vouchers

Selected user questions

Consultation

„I already use ML for my research project, however most of my colleagues work with different techniques.“

We act as a „sparring partner“, give feedback and suggestions, research minor issues, ...

Performance

„I have running ML code but it does not scale well / does not run fast enough.“

We support you with performance analysis and code review

Implementation

„I have data at hand and I know what I want to achieve, but I need help with my ML research project.“

We implement code for ML training, data processing, ...

Your question?

No voucher is like the other – contact us for help with your machine learning project!



ML for Earth System Modelling

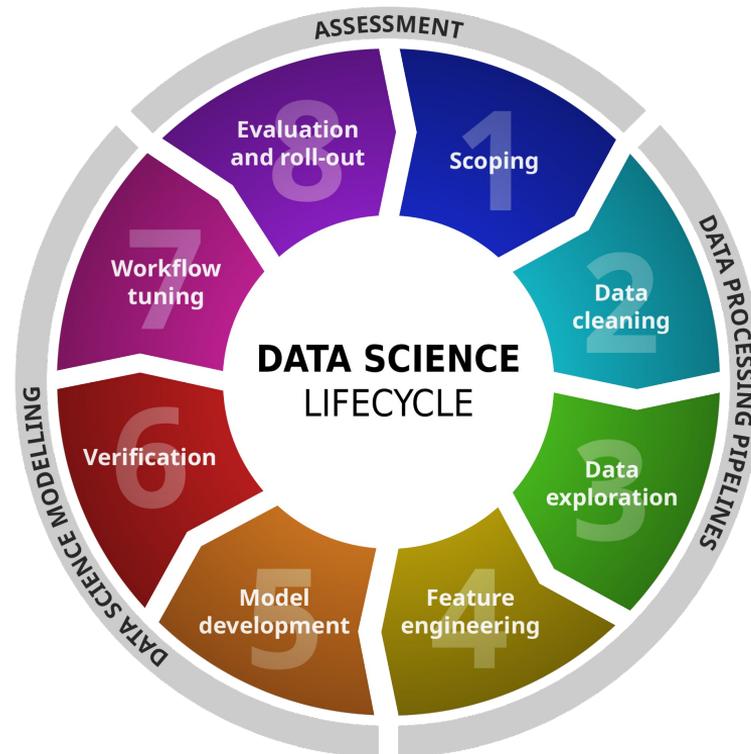
- **Building ESM-ML hybrids**
- Parametrizations for ICON
- Prototype ML models and integration with ESM code
- Interactive digital twins for extrapolation



ML on Earth observation data

- **Support for training, testing, tuning, deploying EO workflows with ML**
- Time series, example: Seismology
- Image-like data (e.g. jpg-images or Satellite images), example: Dead Sea Lakebed Segmentation

Full Data Science Lifecycle



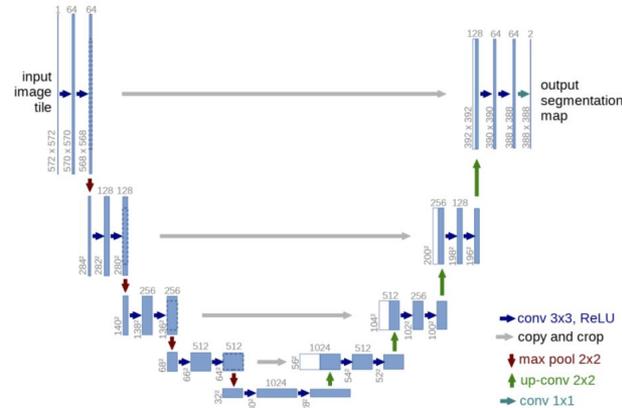
Dead Sea Lakebed Segmentation

Objective:

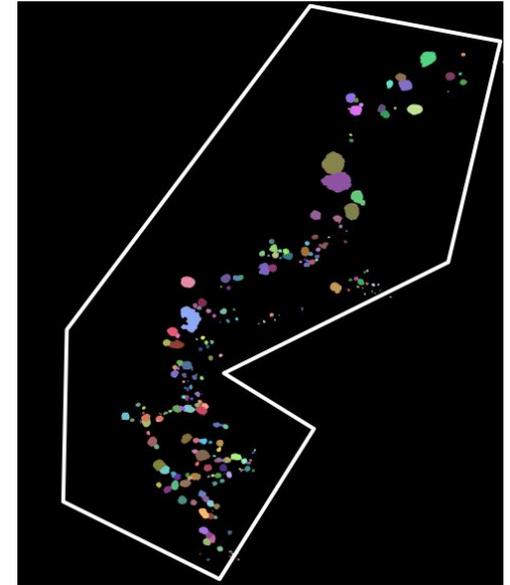
Perform sinkhole **Instance Segmentation** using deep learning techniques



Input



UNet Architecture



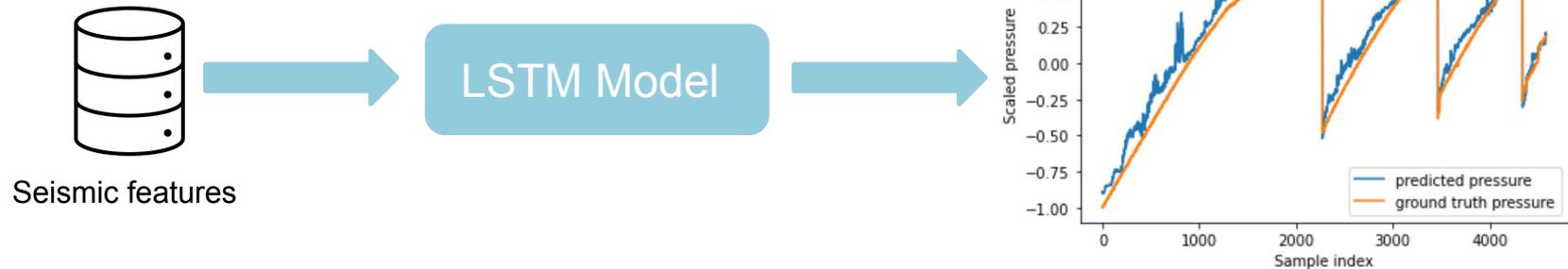
Sinkhole Instance Segmentation

Objective:

Importance of **scientifically established** seismological features through the eyes of a neural network

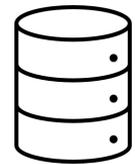
Phase 1:

Train optimal sequence model using a **Long Short Term Memory** neural network to predict pressure



Phase 2:

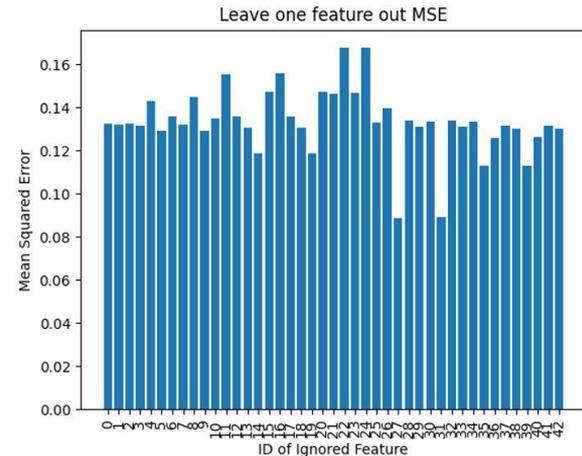
Using **test** data and the **optimized** model from phase 1 to compute the error when one feature is ignored



Test data



Optimized LSTM Model



ML Emulators in Earth System Models (ESMs)

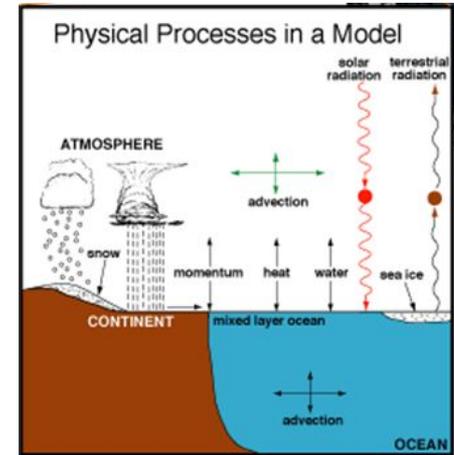
- ESM: few km grid resolution
- Processes on smaller scales, e.g.
 - convection
 - atmospheric chemistry
- Treatment in the ESM
 - parameterization
 - neglected due to computational effort



ICON: ICOSahedral
Nonhydrostatic model

Replace sub-grid scale process by ML prediction

- “Easy” to train offline with good accuracy
- Need to test online - coupled to the ESM

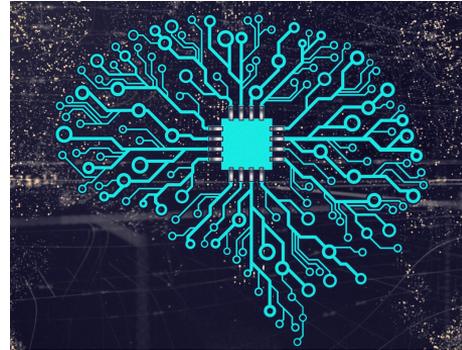


Hybrid Models: Rapid Development

FORTRAN / CPU



Python / GPU or CPU



Voucher request

- Call Python ML inference inside Fortran ESM
- Should allow for iterative development - not hard coded
- Should be performant to run experiments

What is required for a good Voucher Problem?



(Big) data



A challenge



A client

Contact us to jumpstart your AI project:
consultant-helmholtz.ai@dkrz.de
Tobias Weigel, weigel@dkrz.de

<https://www.helmholtz.ai>

<https://docs.dkrz.de/doc/software%26services/machine-learning/index.html>