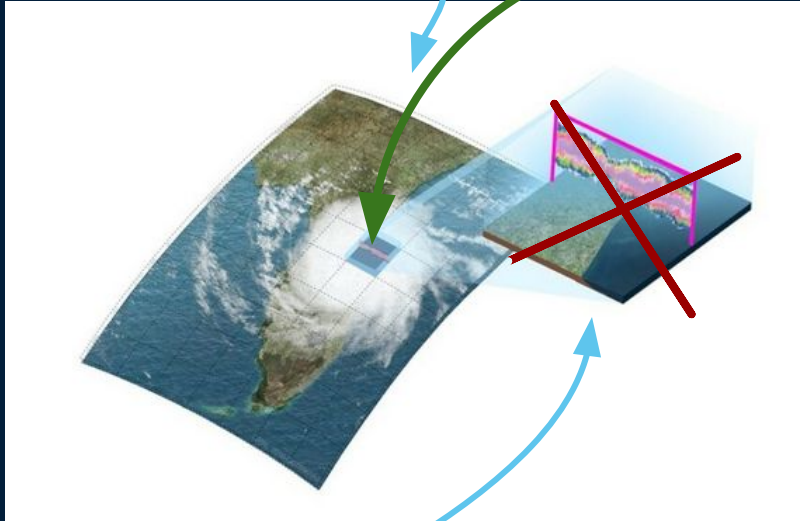

LEAP: Atmospheric Physics using AI

Final Presentation

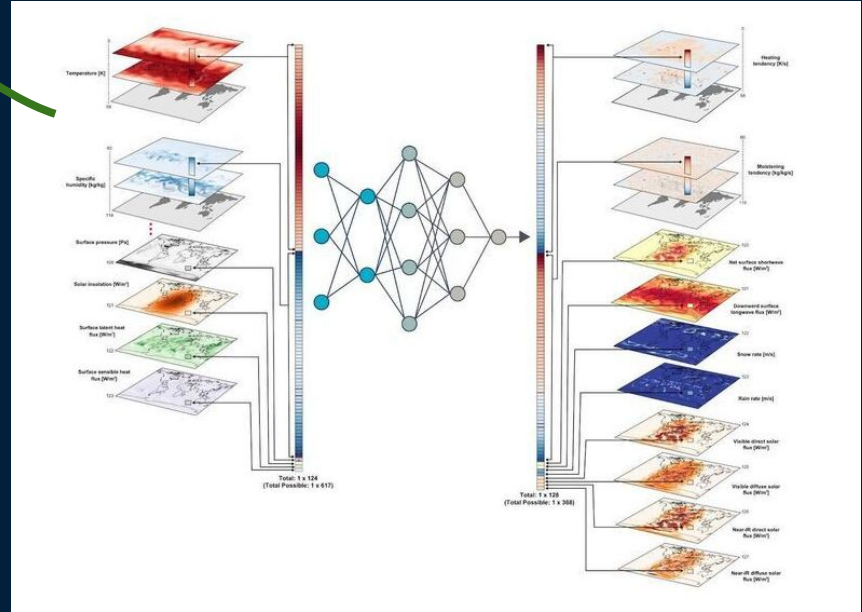
Objective: a data-driven parametrization of small scale processes

numerical model



fine-scale model = expensive!!

-> parametrizations ... not always very accurate



The Approach: Divide in 5 Categories



Temperature:

- heating tendency profile



Clouds:

- moistening tendency profile, liquid clouds & ice clouds



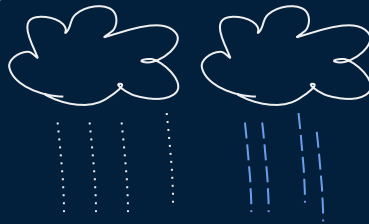
Wind:

- zonal acceleration
- meridional acceleration



Radiation:

- shortwave radiation (5 individual parameters)
- longwave radiation



Precipitation:

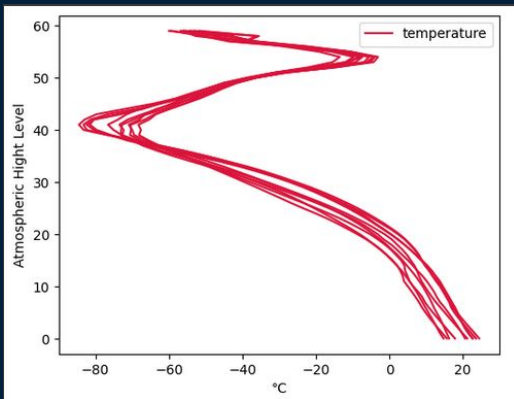
- rain rate & snow rate

Data Pipeline: The Dataset

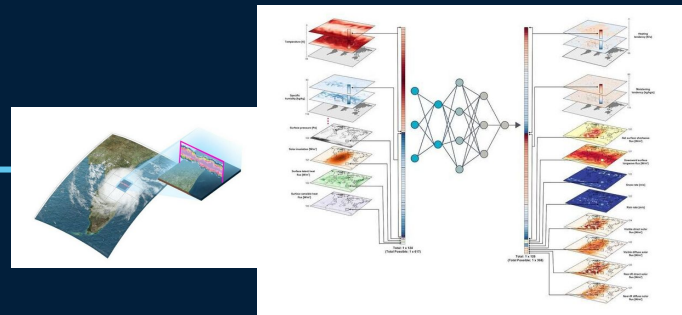
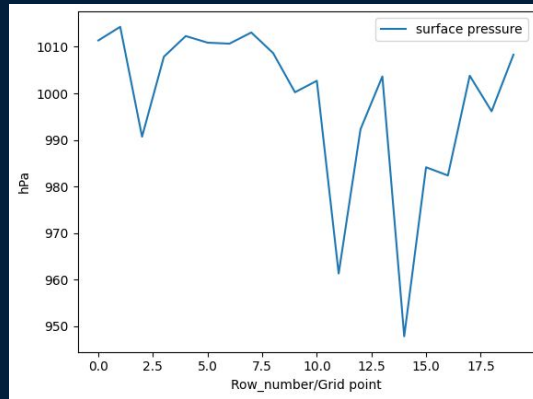
CSV file (~300GB) containing the training data:

- each row represents one point in space and time (~10 MIO rows!)
- 556 columns corresponding to **25 input** variables (from coarse model)
- 368 columns corresponding to **14 target** variables (from fine model)

some inputs/targets are vertical profiles:



others are just “scalars”:



Data Pipeline: Preprocessing

Feature Selection

- > based on physical knowledge and linear correlations
- > a dictionary that maps the relevant input feature to the targets

Normalization

- > column wise to not mix up different units
- > scaling with min and max to $[-1,1]$

Formatting & Splitting

- > 1D feature tensors and 1D target tensors of varying length
- > 100000 row train set, 80% train, 20% validation
- > 1000 row test set

A Family of Models:

train an individual model for each (sub)category:

- Temperature
- Clouds
- Wind (2 models)
- Radiation (2 models)
- Precipitation

baseline: simple MLP (two hidden layers,

ReLU,

number of neurons depend on input size)

A Family of Models:

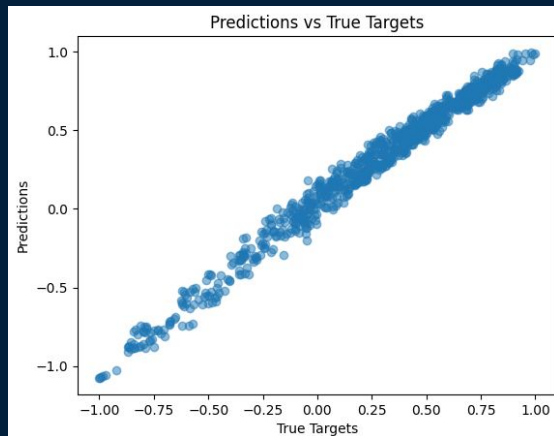
-> some targets have simple relationships to their input features:

-> a simple MLP is sufficient to get good predictions

example: longwave radiation

R^2 score = 0.98

$$R^2 = 1 - \frac{\sum (y_{\text{true}} - y_{\text{pred}})^2}{\sum (y_{\text{true}} - \bar{y}_{\text{true}})^2}$$



The figure shows (normalized) predictions of the MLP vs (normalized) true targets of a test set (1000 rows).

A Family of Models:

-> other targets have more complicated relationships:

-> need more sophisticated architecture than the simple MLP (e.g. transformer)

example: heating tendency profiles

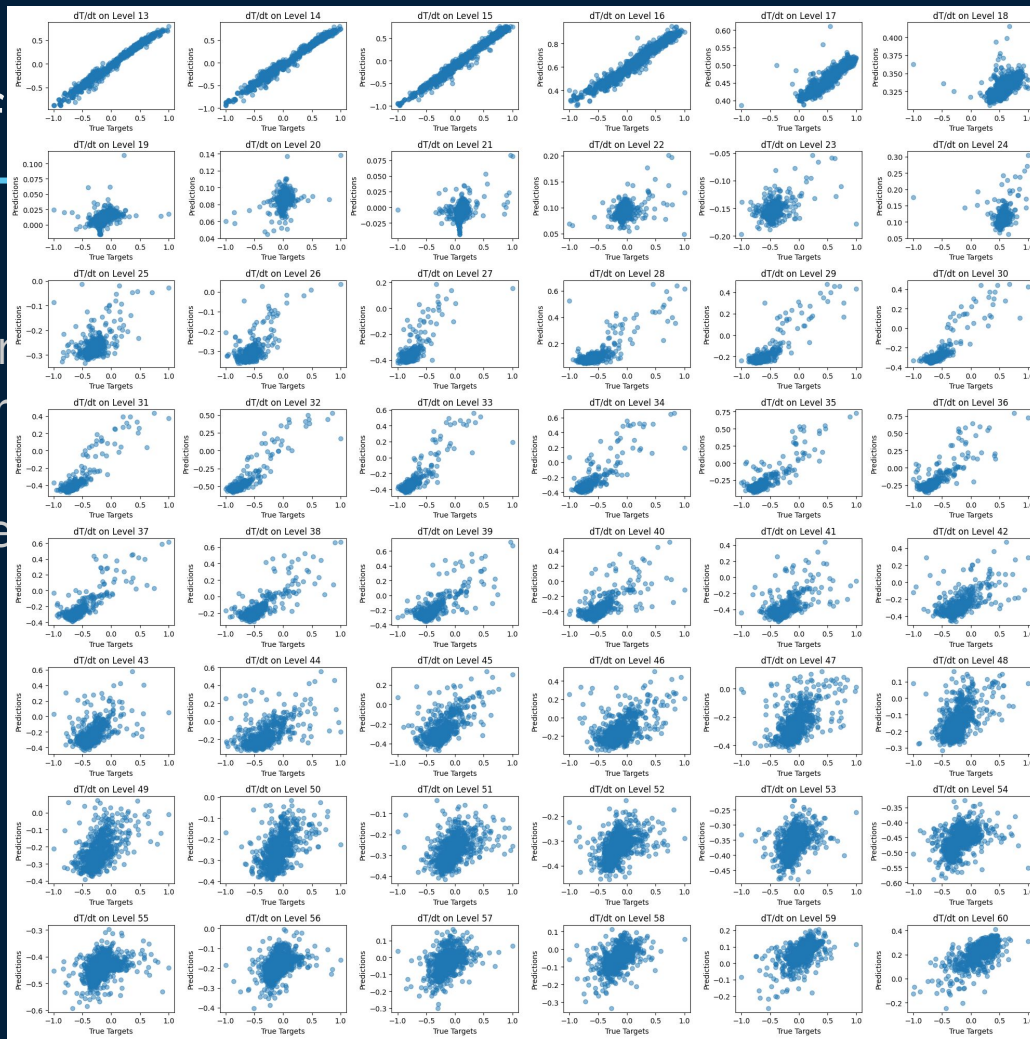
A Family of

-> other targets

-> need more

transformer

example: he



MLP (e.g.

Outlook for Remaining Time:

- > train a transformer model for the more complicated targets
- > explore if already well predictable targets can help predicting others
- > train as many models as possible on the full dataset

Summary in 5 Categories



Temperature:

- heating tendency profile



Clouds:

- moistening tendency profile, liquid clouds & ice clouds



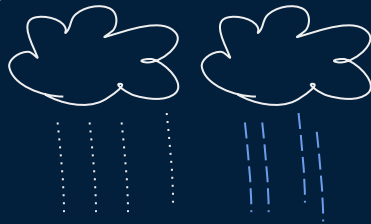
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Precipitation:

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