

Development and Evaluation of Coupled Climate Simulations Using Machine Learning-Enhanced Aerosol Model

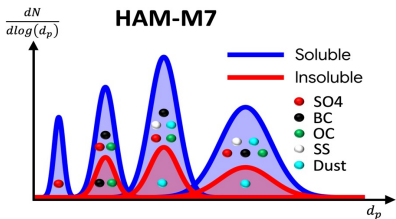
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1. Background

Global-Scale Aerosol Modelling

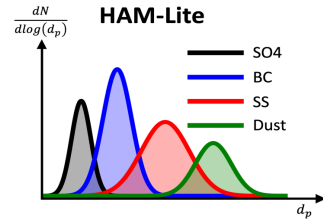
- TerraDT aims to advance Earth's digital twins as part of the Destination Earth initiative.
- Microscopic aerosol processes are complex and computationally demanding to model over large areas, but essential for representing their kilometer-scale impacts on cloud systems.
- Current aerosol models are either computationally too heavy or too simplified and inaccurate for global scale modelling.



1. Complex Aerosol Model HAM-M7

- Simulates aerosols as 7 log-normal modes and total of 25 tracers.
- Tracers can be soluble or insoluble and are divided into different size classes.

The model is computationally too heavy for global scale modelling.



2. Simple Aerosol Model HAM-Lite

- Simplified version of HAM-M7. Simulates aerosols as 4 tracers, each in one soluble log-normal mode.
- The model includes emissions and sinks, but not any microphysics between them.
- Computationally efficient for global scale modelling.

The model is too inaccurate.

2. Problem

Too Simplified Model

- HAM-Lite is made for global scale modelling but it is too simplified and inaccurate to make good predictions. It has present, constant hygroscopicity for each tracer and very simplified calculation for extinction coefficient.
- With more detailed processes the model would become computationally too heavy. It needs to be made more accurate without increasing the computational cost significantly.

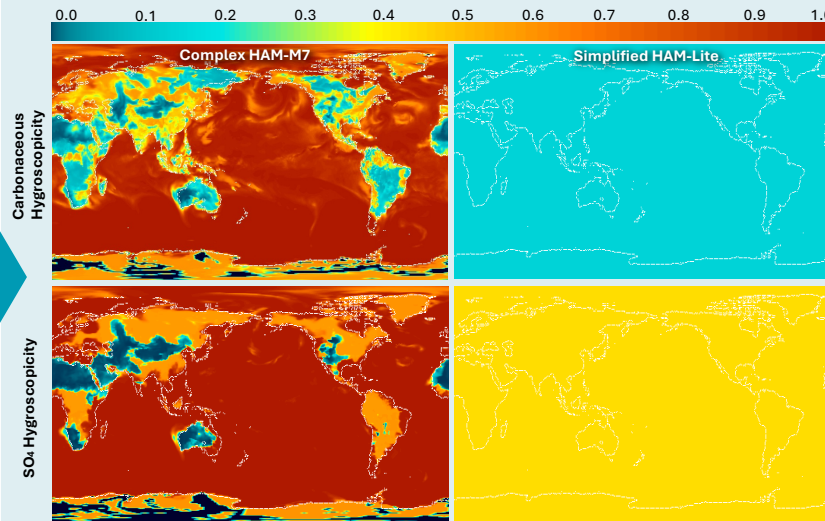


Figure: Carbonaceous and SO₄ hygroscopicity from HAM-M7 (left) and HAM-Lite (right).

3. Solution

Boosting with Machine Learning

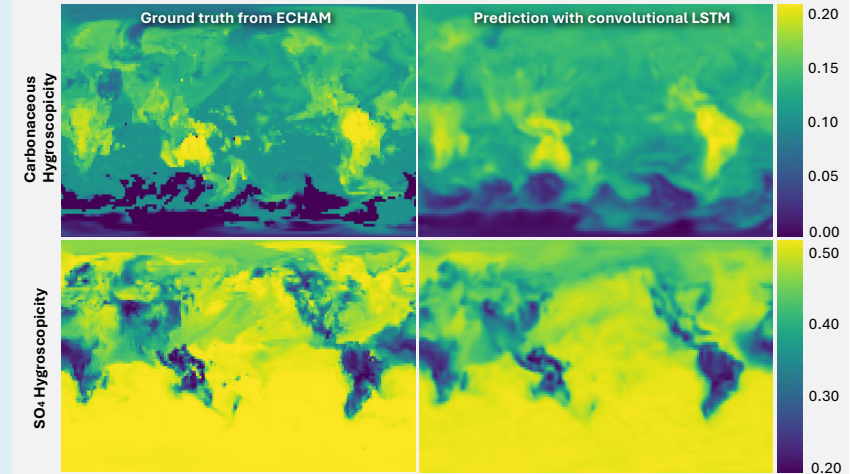
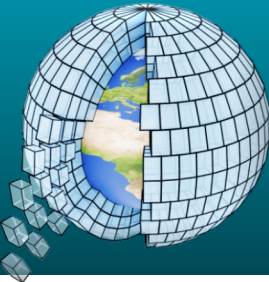


Figure: First predictions of hygroscopicity at surface level using Convolutional LSTM and training data from ECHAM. Carbonaceous hygroscopicity (top) and SO₄ hygroscopicity (bottom) from ECHAM and corresponding predictions.

- Machine learning component is implemented for HAM-Lite to predict hygroscopicity κ and extinction coefficient α_{ext} .
- The training data is obtained from the complex HAM-M7 aerosol model coupled with a global climate model Open-IFS. Meteorological values and tracer concentrations from surrounding grid points in space and time are used as input data.

5. Take-Home Message

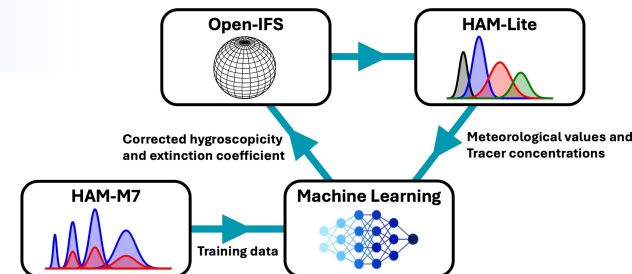
- Current aerosol models are either too simplified or computationally too heavy for global scale climate modelling.
- In this study we enhance a simple and lightweight aerosol model HAM-Lite with machine learning methods. Training data is obtained from a more accurate and computationally heavy aerosol model HAM-M7.
- The new aerosol model learns to predict results of a more complex model with less computational cost. This model will be coupled with a global climate model.



4. Goal

Coupled System

- Goal is to have a lightweight and accurate aerosol model with a machine learning component correcting hygroscopicity and extinction coefficient at runtime.
- Simulations will be implemented and evaluated with a global kilometer-scale climate model Open-IFS and later with Integrated Forecasting System (IFS), coupled with the new boosted aerosol model.
- IFS coupled with boosted HAM-Lite will be the atmospheric component of TerraDT digital twins.



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