

# Neural Representation of the Stratospheric Ozone Chemistry

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# Overview: Stratospheric Ozone Chemistry in Climate Science

Lagrangian Chemistry &  
Transport Model ATLAS

- ☒ Computation time
- ✓ Accuracy



Neural-SWIFT

- ✓ Computation time
- ✓ Accuracy

Stratospheric Chemistry

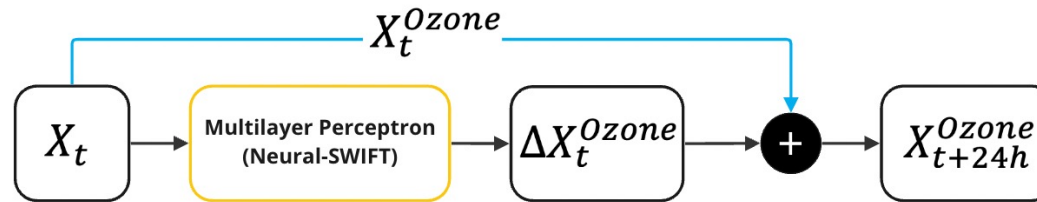
- ~ 47 chemical species
- ~ 171 chemical reactions

Prescribed Ozone tables

- ✓ Computation time
- ☒ Accuracy

# How does Neural-SWIFT work?

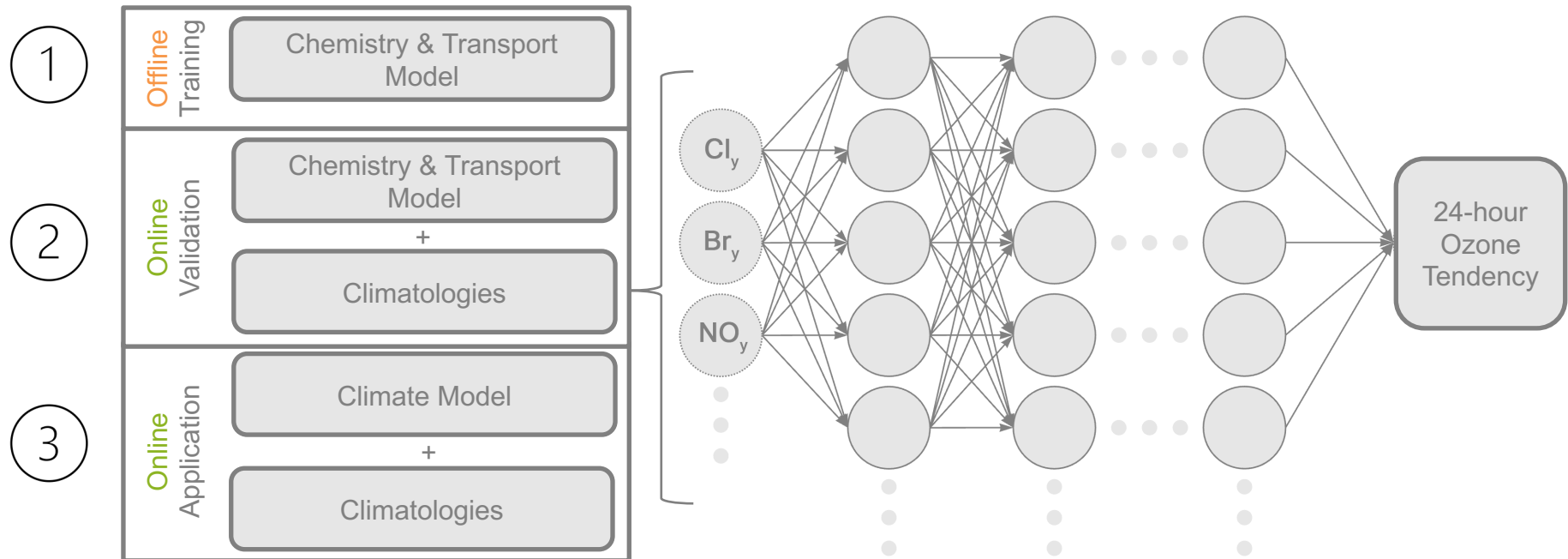
Applying Multilayer Perceptrons in the forward Euler scheme



where  $X_t$  are input variables at time  $t$ ;  
 $X_t^{Ozone}$  is the ozone variable at time  $t$  [volume mixing ratio];  
 $\Delta X_t^{Ozone}$  is the 24-hour ozone tendency;  
 $X_{t+24h}^{Ozone}$  is the updated ozone variable;

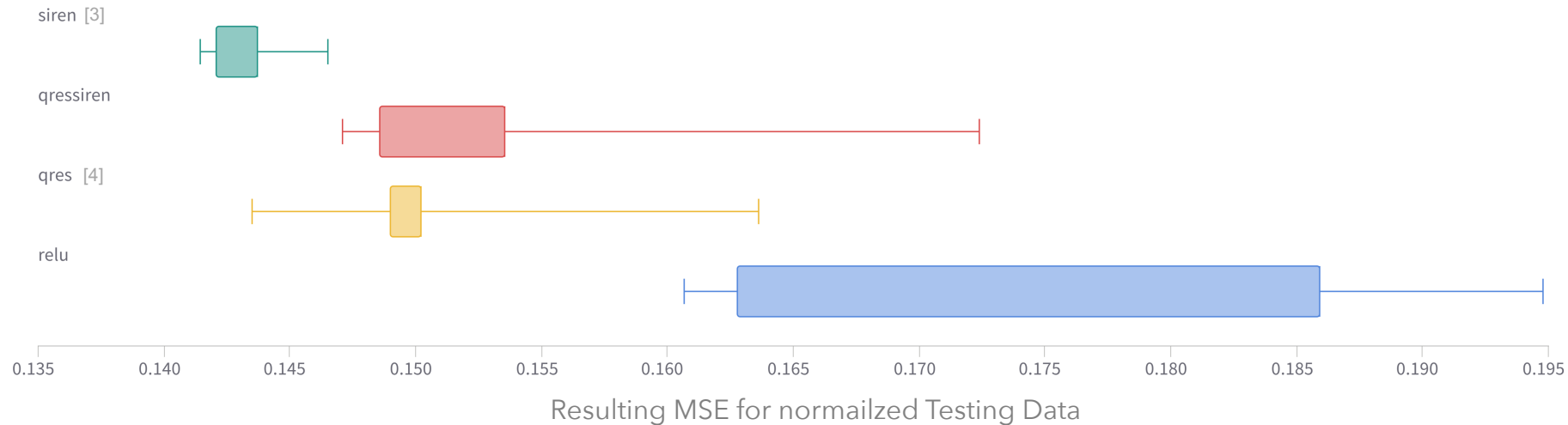
# From Input-Output Data Pairs to Application

What sources of input variables (point-wise) can we consider?



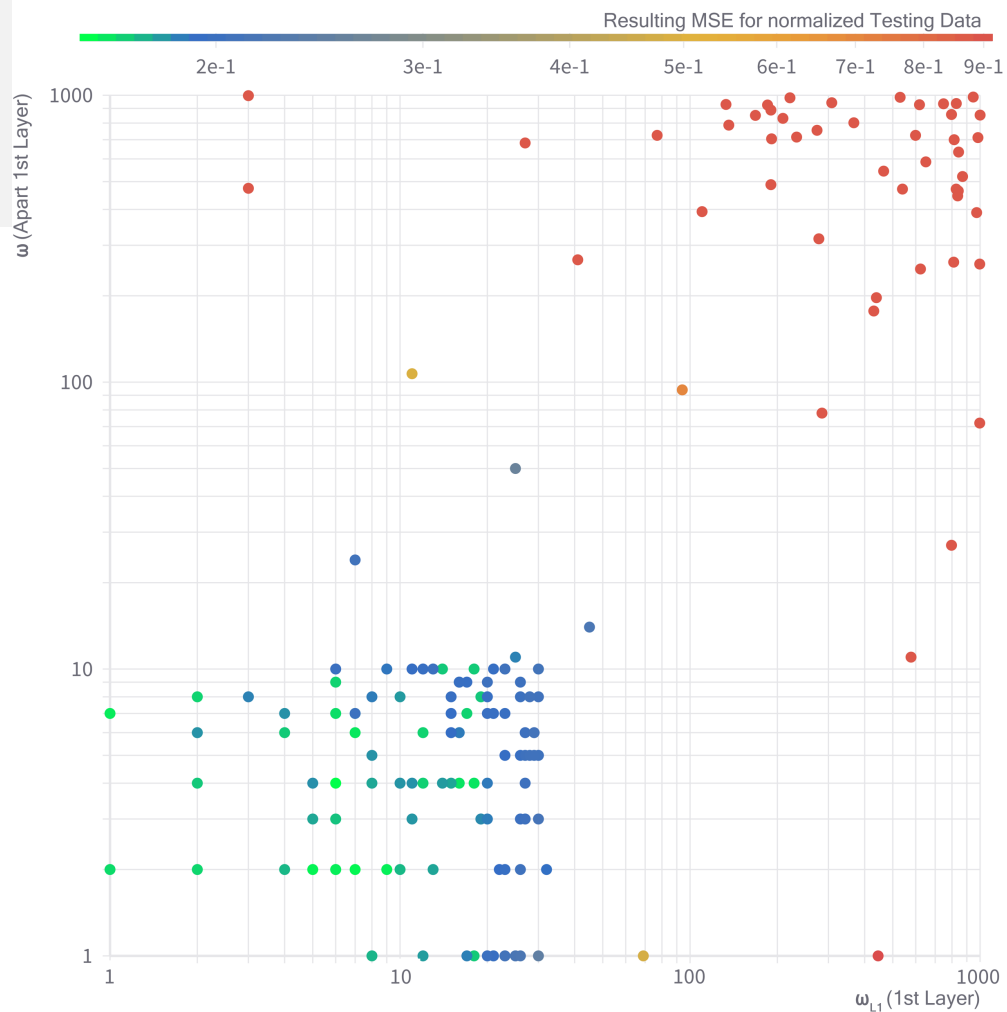
# Architectures of Multilayer Perceptrons

Which available architecture performs best on our data set?



# Hyperparameter Optimization

- Bayesian Hyperparameter Search
- Hyperband Early Stopping [5]
- Tuned Hyperparameters:
  1. Number of layers
  2. Number of neurons per layer
  3. Siren [3] parameter  $\omega_{L1}$  for first
  4. ... and one  $\omega$  for other layers
  5. Learning rate
  6. Mini-Batch size



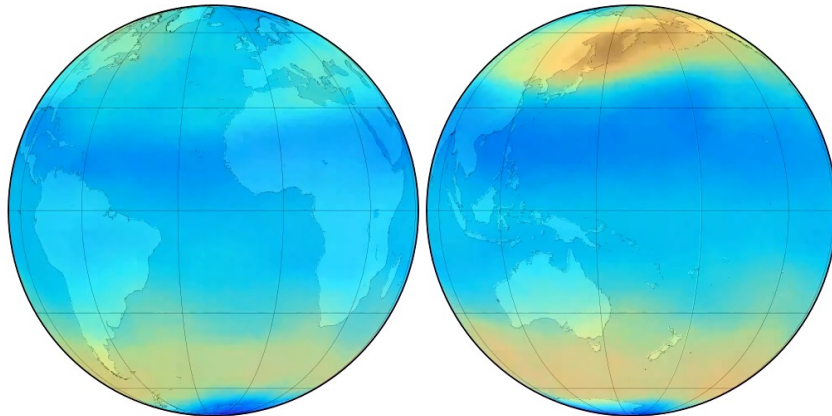
# Accuracy & Stability

Is there an error accumulation?

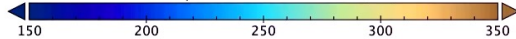
Two-year simulation of Neural-SWIFT coupled to ATLAS:

## Neural-SWIFT

Time coordinate values: 1998-11-30 00:00



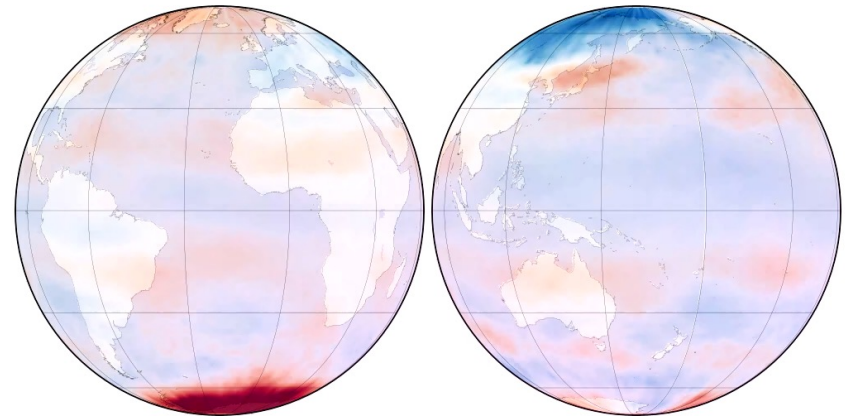
Stratospheric Ozone Column (DU)



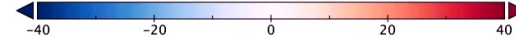
Data Min = 186, Max = 336, Mean = 258

## Neural-SWIFT - Full Chemistry

Time coordinate values: 1998-11-30 00:00 : Time coordinate values: 1998-11-30 00:00



Stratospheric Ozone Column (DU)



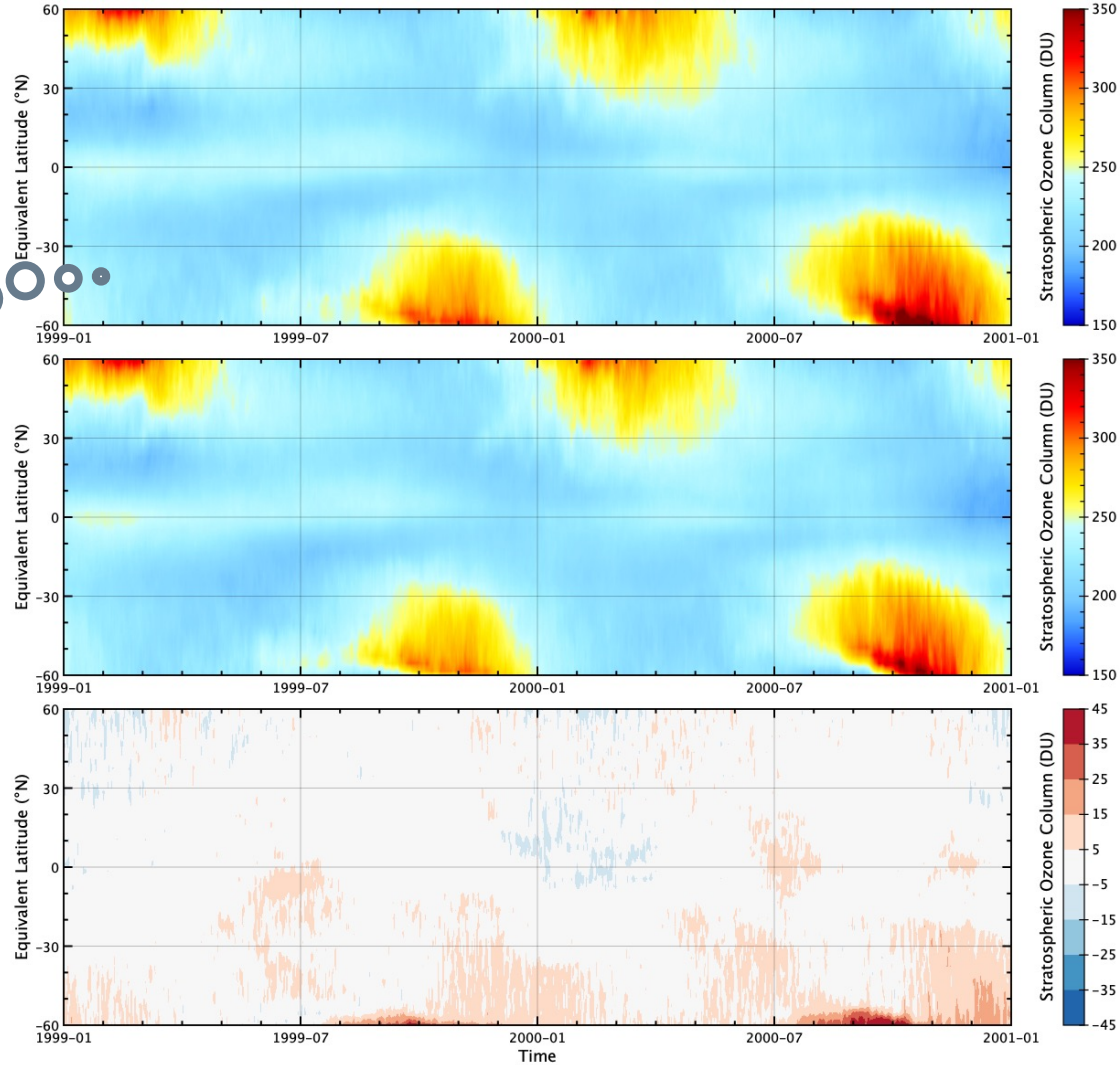
Data Min = -39, Max = 56, Mean = 1

Neural-SWIFT

Accurate & stable  
over time

Full Stratospheric Chemistry

Difference:  
[Neural-SWIFT] – [Full chemistry]





# Speed-Up

Are AI surrogate models significantly faster?



~700 x faster

Simulation	Neural-SWIFT	Full Chemistry <sup>[2]</sup>
1 model-day	3.4 s	40 min
1 model-year	21 min	> 10 days
100 model-years	< 1.5 days	> 1000 days

Note: Neural-SWIFT was coupled to the chemistry and transport model ATLAS and run on the same server with 48 CPUs 1.0-3.9 GHz, 755 GB physical memory. Calculation time refers to chemistry calculation only and does not include time required for transport and mixing in ATLAS.

# Key takeaways

- **Neural-SWIFT** is a data-driven model that enables an **interactive** representation of the **stratospheric ozone chemistry** and is intended for application in climate models.
- Benchmark Dataset open available: <https://doi.org/10.1594/PANGAEA.939121>
- Neural-SWIFT achieves:
  - ✓ Speed-Up: **factor of 700**
  - ✓ Very accurate
  - ✓ Interactive
  - ✓ Stable: No **significant error accumulation** within a 2-year simulation

# Sources

1. <https://www.esrl.noaa.gov/csl/assessments/ozone/2018/downloads/twentyquestions/>
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3. Sitzmann, Vincent, Julien N. P. Martel, Alexander W. Bergman, David B. Lindell, and Gordon Wetzstein. "Implicit Neural Representations with Periodic Activation Functions." ArXiv:2006.09661 [Cs, Eess], June 17, 2020.
4. Bu, Jie, and Anuj Karpatne. "Quadratic Residual Networks: A New Class of Neural Networks for Solving Forward and Inverse Problems in Physics Involving PDEs." ArXiv:2101.08366 [Cs], January 27, 2021.  
<http://arxiv.org/abs/2101.08366>.
5. Li, L., Jamieson, K., DeSalvo, G., Rostamizadeh, A., Talwalkar, A., 2018. Hyperband: A Novel Bandit-Based Approach to Hyperparameter Optimization. arXiv:1603.06560 [cs, stat].
6. Mohn, Helge; Kreyling, Daniel; Wohltmann, Ingo; Lehmann, Ralph; Rex, Markus (2021): Benchmark dataset for 24-hour stratospheric ozone tendencies (SWIFT-AI-DS). PANGAEA, <https://doi.org/10.1594/PANGAEA.939121>