



# CAMulator: Fast Emulation of the Community Atmosphere Model

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## Our Framework: CREDIT

### What is CREDIT?

An open foundational platform for developing and deploying AI weather and Earth system prediction models.

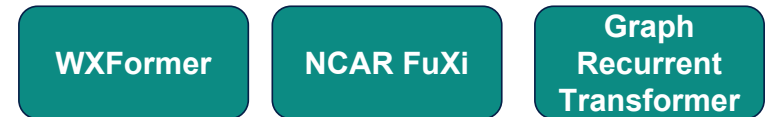
CREDIT enables users to build custom data and modeling pipelines to load data, train configurable AI forward models, and deploy them for real-time forecasting, hindcasting, or scenario projections.

CREDIT offers both scientifically validated model configurations and endless customization for any use case.

#### Datasets



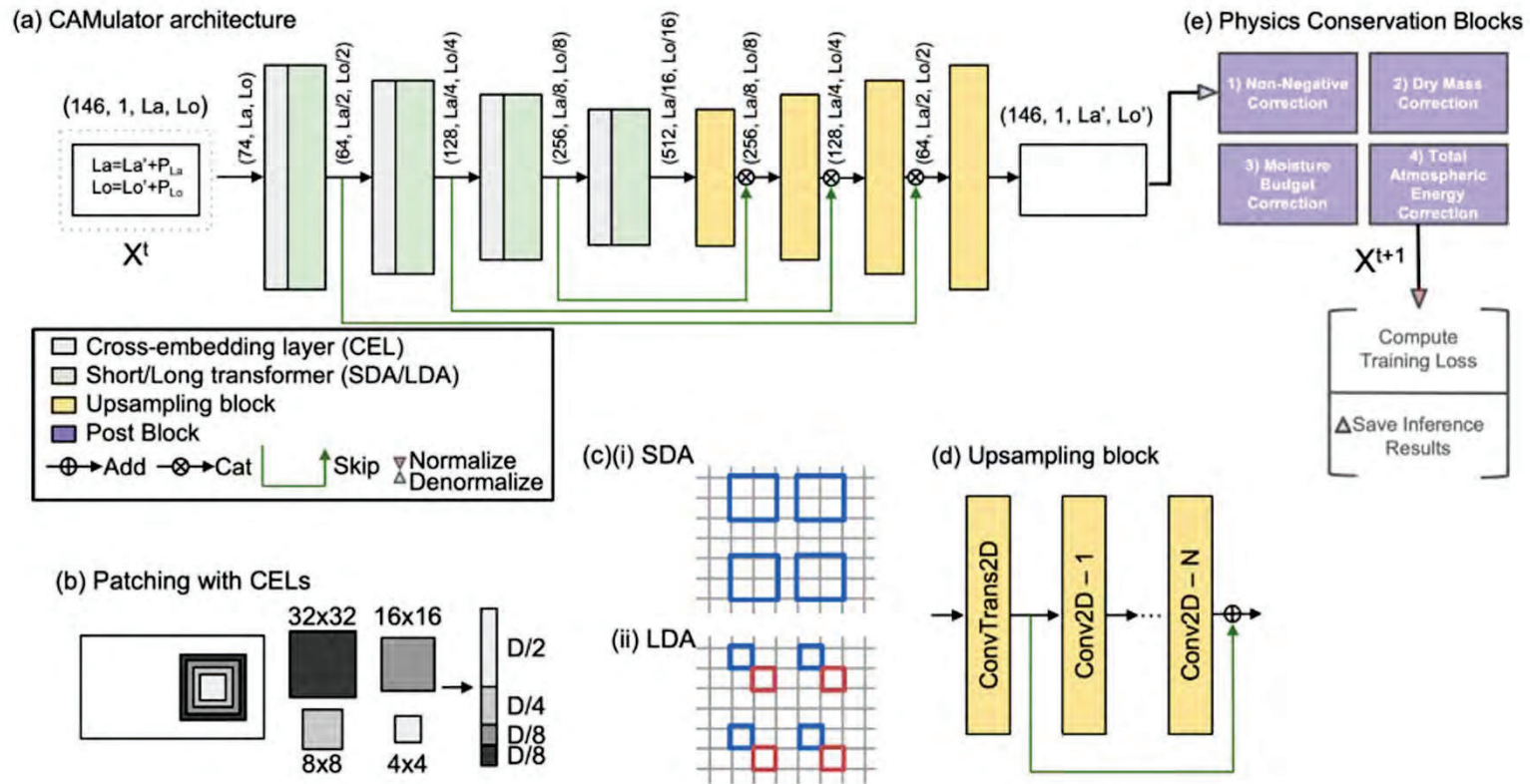
#### Models



#### Physics



# Variables and Architecture



- Trained on CAM over 30 year observed record forcing
  - Throughput of 480 Sim. Years/Day

# FHIST Run

## Data Prep Workflow

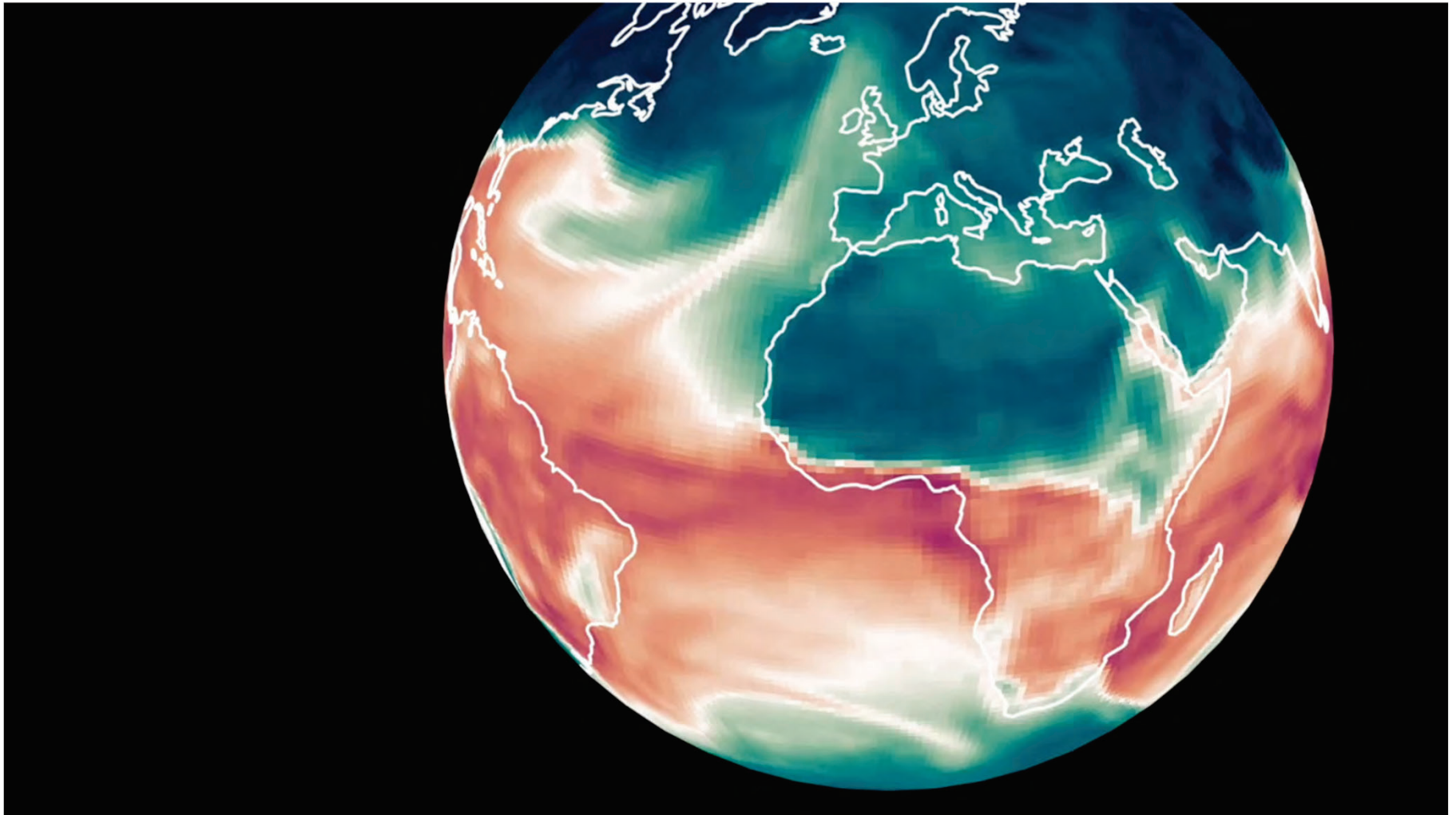
- 1) Conduct a 35-year FHIST run (1979-2014) of **CAM6 in CESMv2.1.5**
- 2) Collect 6-hourly data
- 3) Compute Total Water + Convert to Flux forms
- 4) Gather Static & Forcing Data

## WE ARE GOING TO EMULATE THESE VARIABLES:

Variable	Description	Units	Single Level/Levels	I/O
<b>Prognostic Variables (Input and Output)</b>				
U	Zonal Wind	m/s	32 levels	Input/Output
V	Meridional Wind	m/s	32 levels	Input/Output
T	Temperature	K	32 levels	Input/Output
Qtot	Specific Total Water	kg/kg	32 levels	Input/Output
<b>Diagnostic Variables (Output Only)</b>				
PREC	Precipitation Rate	m	Single Level	Output
CLDTOT	Total Cloud Cover	fraction	Single Level	Output
CLDHGH	High Cloud Cover	fraction	Single Level	Output
CLDLLOW	Low Cloud Cover	fraction	Single Level	Output
CLDMED	Medium Cloud Cover	fraction	Single Level	Output
TAUX	Zonal Wind Stress	N/m <sup>2</sup>	Single Level	Output
TAUY	Meridional Wind Stress	N/m <sup>2</sup>	Single Level	Output
U10	10m Wind Speed	m/s	Single Level	Output
QFLX	Surface Moisture Flux	m	Single Level	Output
FSNS	Net Solar Flux at Surface	J/m <sup>2</sup>	Single Level	Output
FLNS	Net Longwave Flux at Surface	J/m <sup>2</sup>	Single Level	Output
FSNT	Net Solar Flux at TOA	J/m <sup>2</sup>	Single Level	Output
FLNT	Net Longwave Flux at TOA	J/m <sup>2</sup>	Single Level	Output
SHFLX	Sensible Heat Flux	J/m <sup>2</sup>	Single Level	Output
LHFLX	Latent Heat Flux	J/m <sup>2</sup>	Single Level	Output
<b>Surface Variables Prognostic (Input and Output)</b>				
PS	Surface Pressure	Pa	Single Level	Input/Output
TREFHT	Near-Surface Air Temperature	K	Single Level	Input/Output
<b>Dynamic Forcing Variables (Input Only)</b>				
SOLIN	Incoming Solar Radiation	J/m <sup>2</sup>	Single Level	Input
SST	Sea Surface Temperature	K	Single Level	Input
<b>Static Forcing Variables (Input Only)</b>				
Surface Geop.	Normalized Surface Height	m <sup>2</sup> /s <sup>2</sup>	Single Level	Input
Land-Sea Mask	Land Mask × Cosine Latitude	unitless	Single Level	Input

WE FORCE THE MODEL WITH THESE VARIABLES (AMIP)

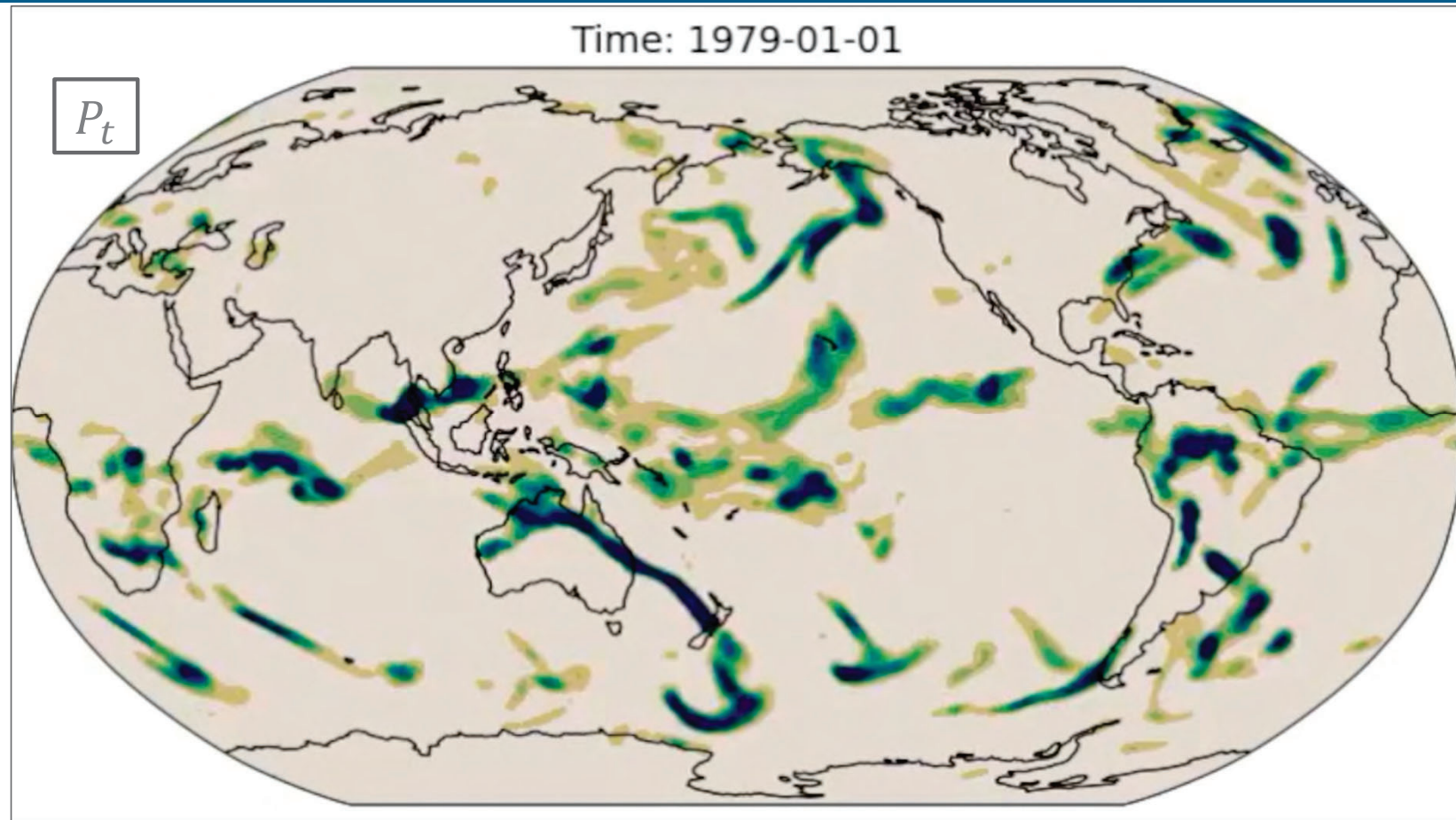
## CAMulatorV1.0



Chapman, Will, et al. "CAMulator: Fast Emulation of the Community Atmosphere ..." *arXiv preprint arXiv:2504.06007* (2025).

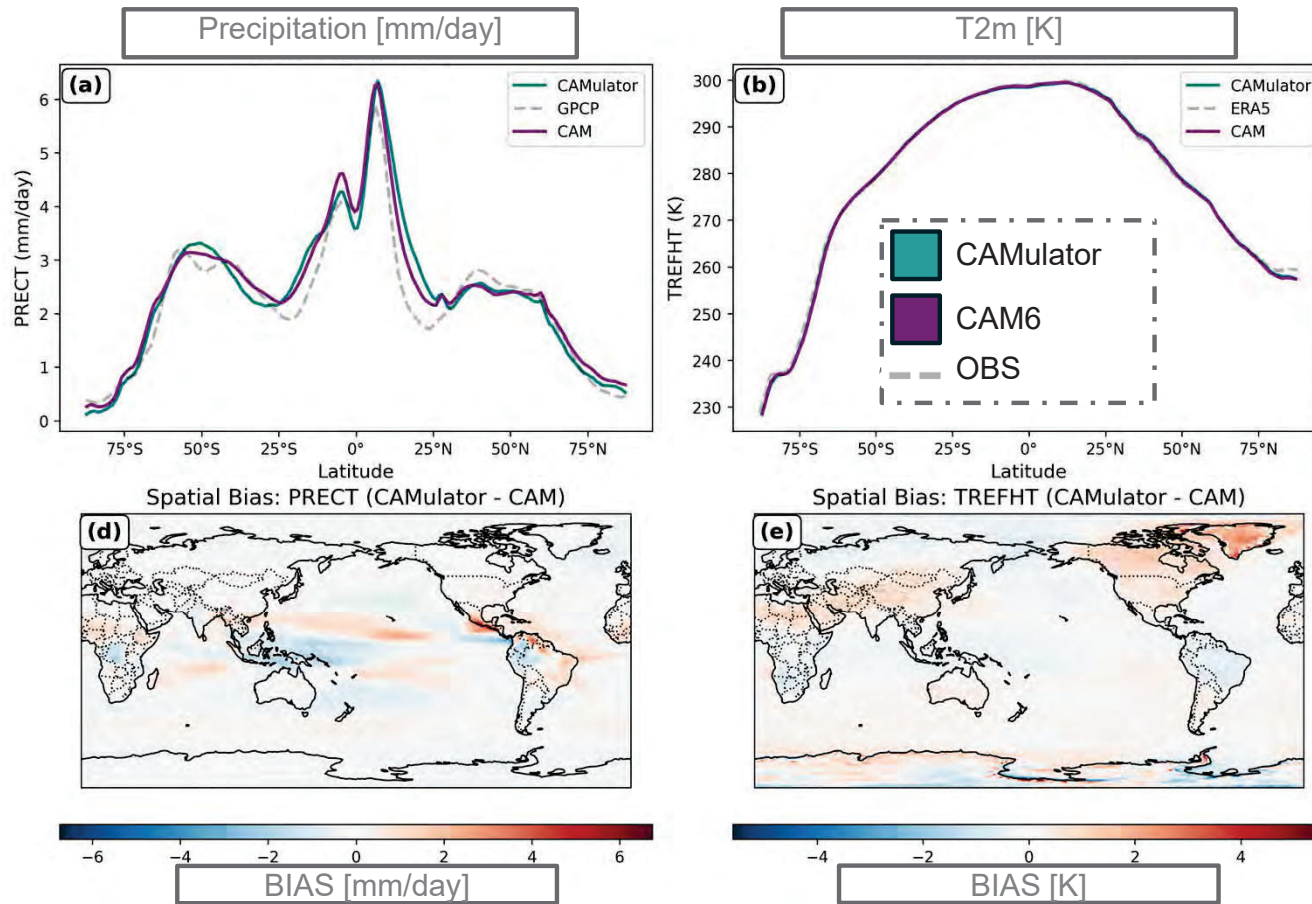


## Precipitation Climatology V 1.0

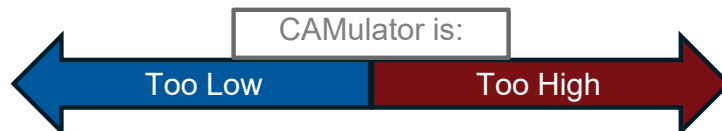


$$P_t = \frac{1}{t} \sum_{i=1}^t \text{PRECIPITATION}(i)$$

# Climatological Mean Bias Error: CAMulator V.1.0



How are the model biases?

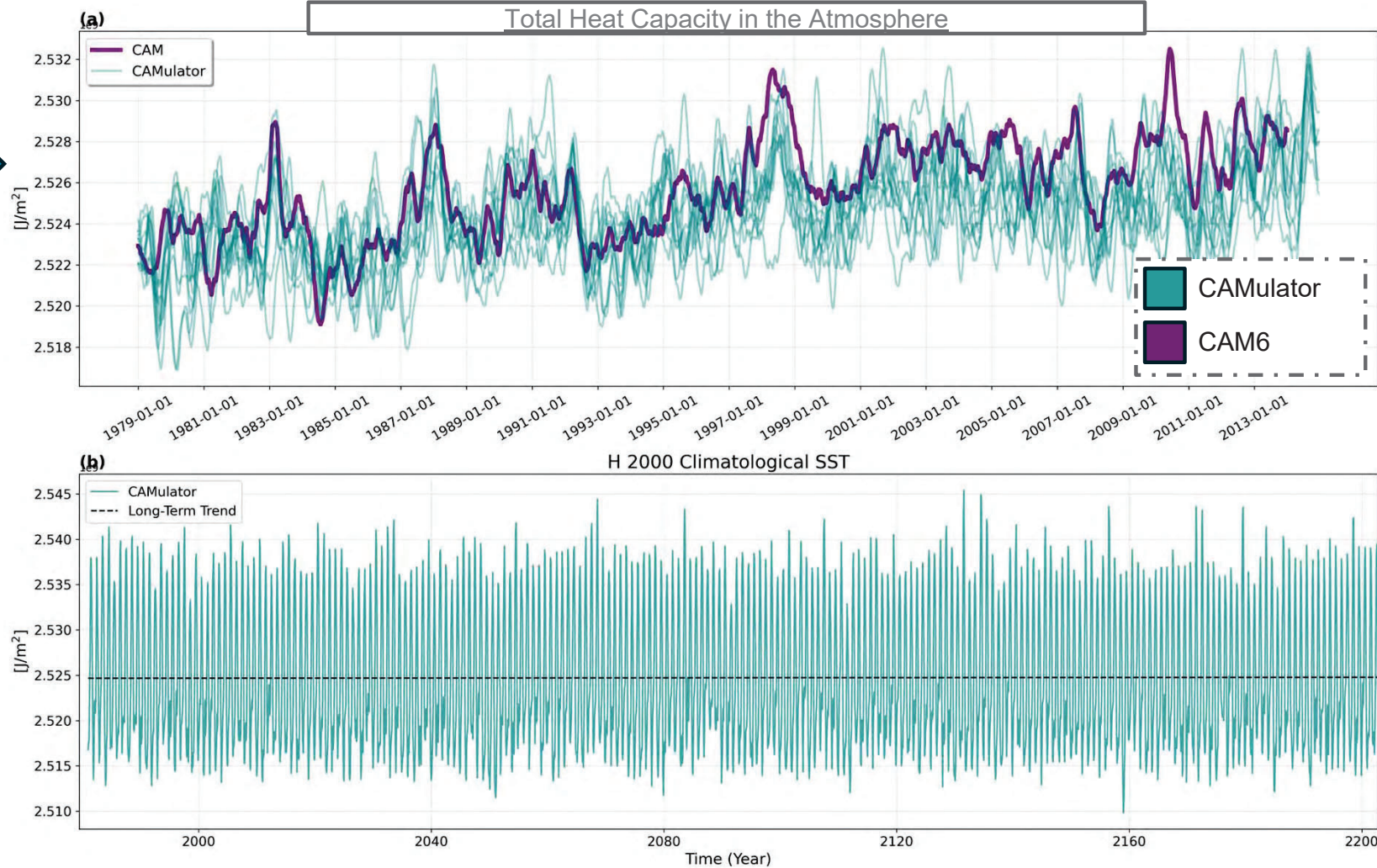


# CAMulator V.1.0

Response to  
historical SST  
forcing

**24 Member 35-  
year ensemble  
in 8 Hours vs.  
4000 Hours for  
CAM6**

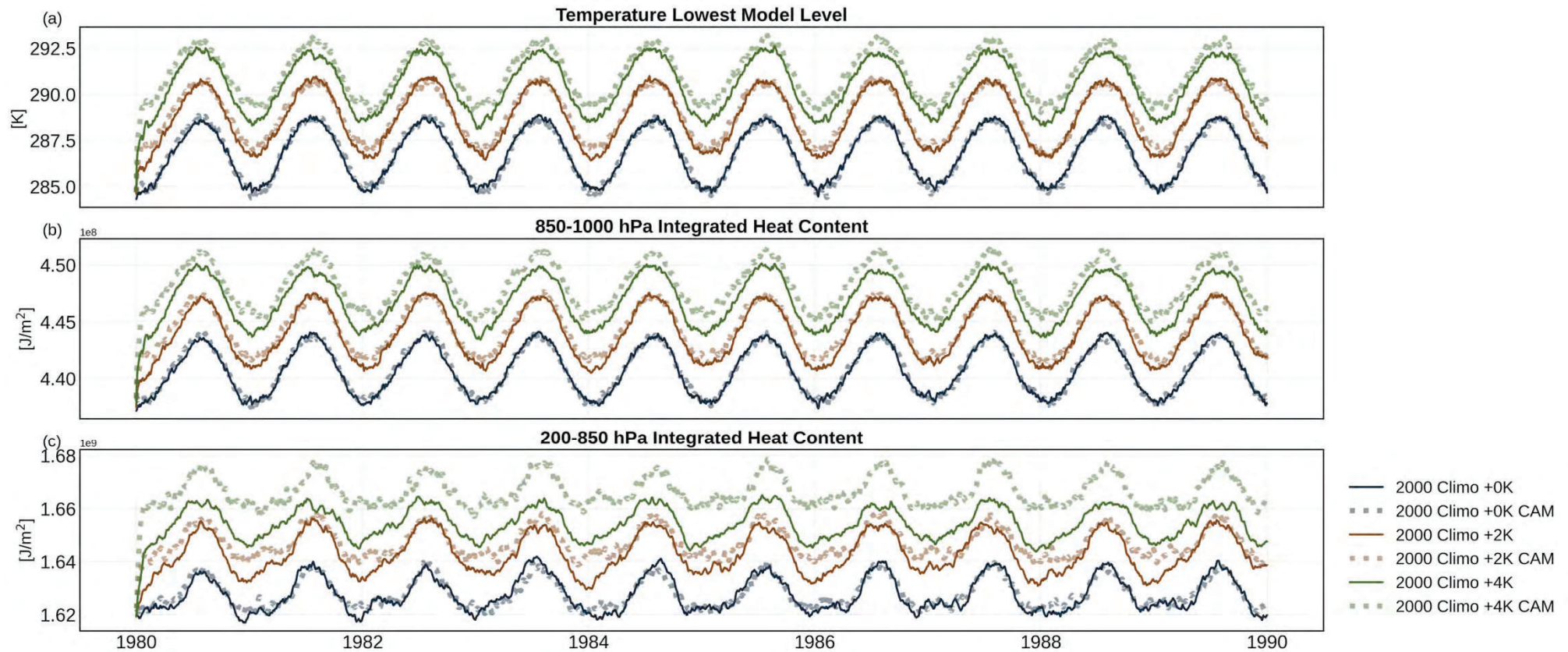
Capable of indefinite roll-  
outs with climatological  
forcing



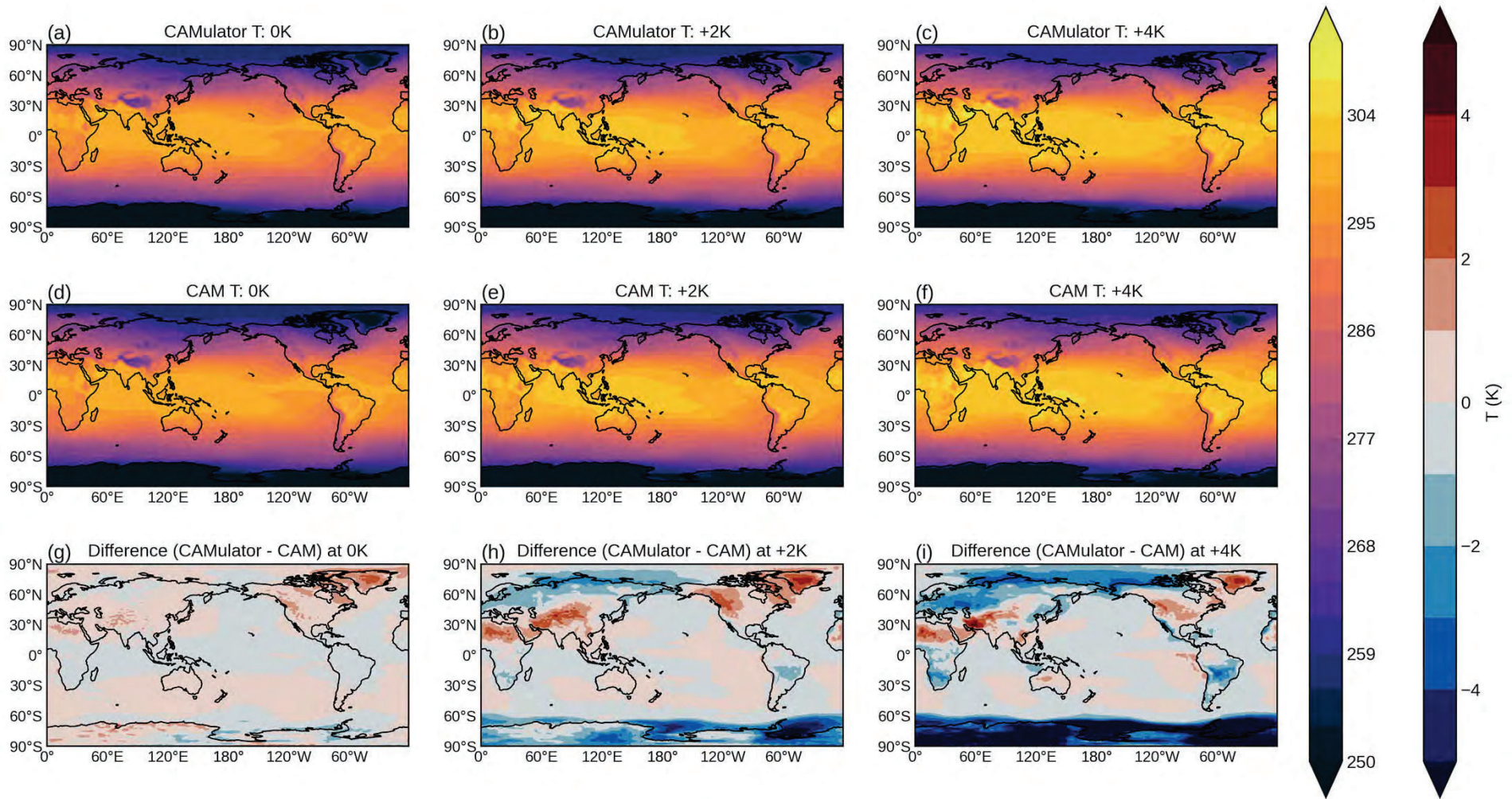
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## Challenge: CAMulator V1.0 SST Forcing Experiments



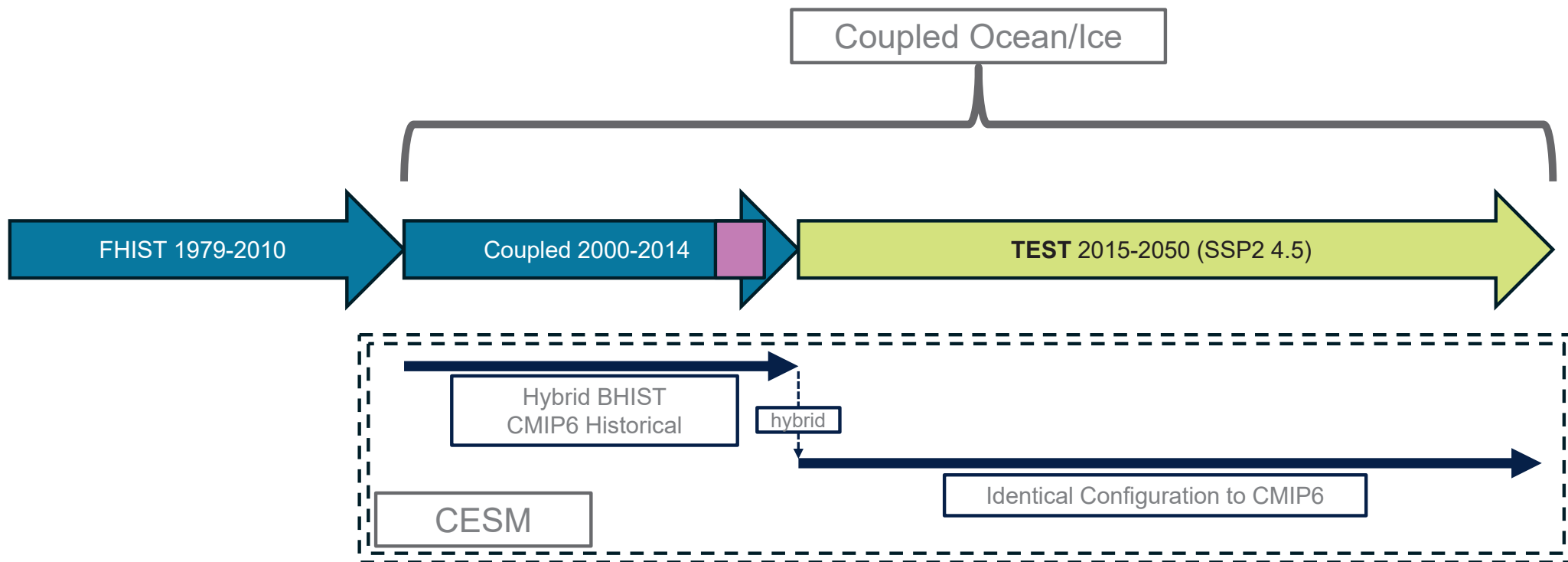
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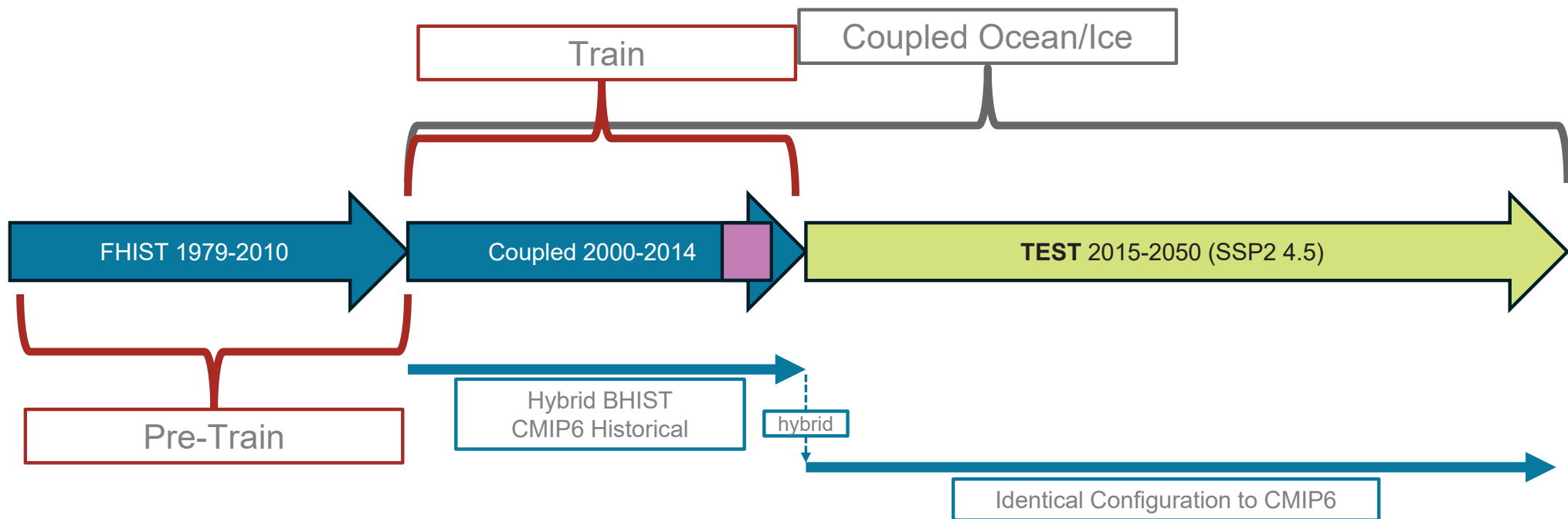
## CESM Data + Training Protocol; Updates:

We are now combining FHIST & Coupled Runs in CESM2.1.5:



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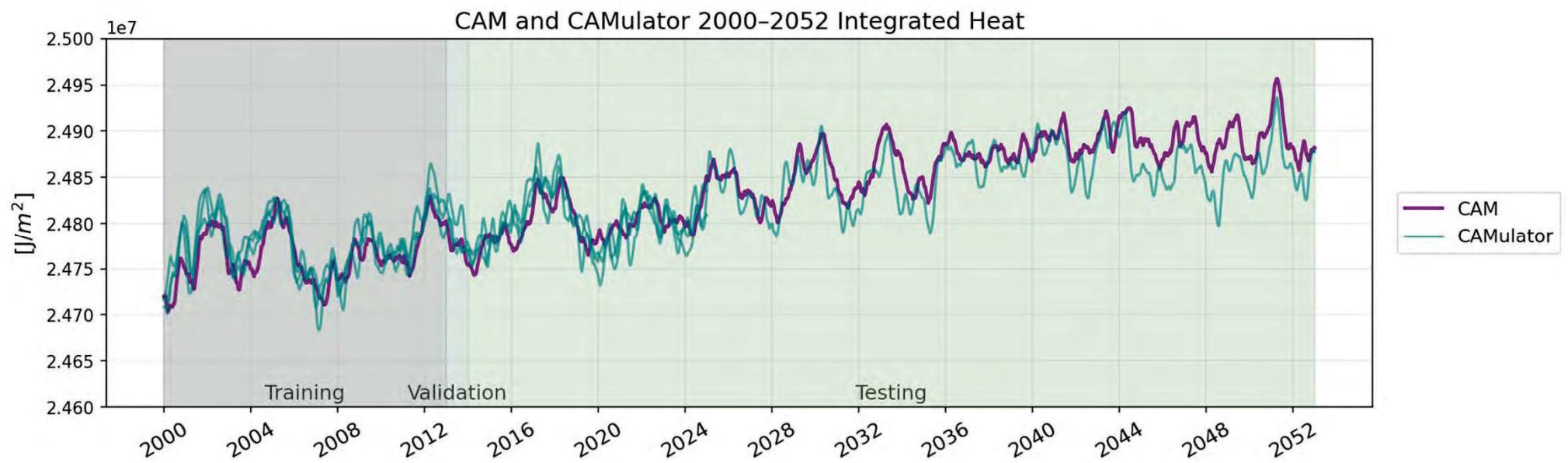




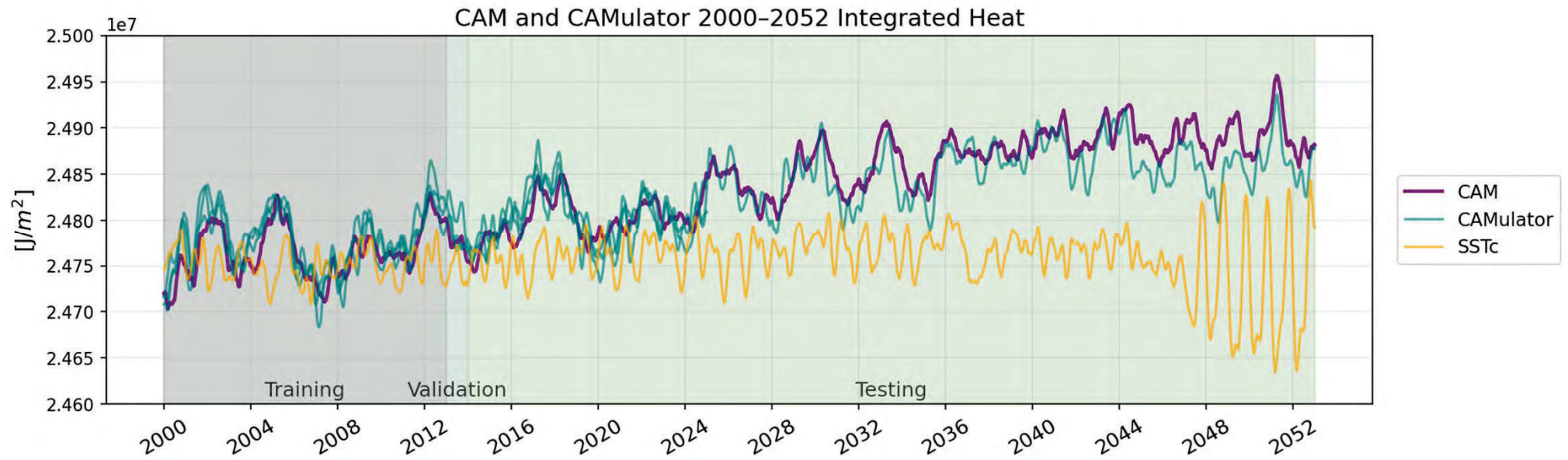
Additional List of Changes from V.1.0:

1. Add Surface Temperature (TS)
2. Add Sea-Ice Fraction
3. Add CO2 Volume Mixing Ratio
4. Add Forcing from Coupled Ocean
5. Remove Redundant Information Variables
6. Remove UpConvTranspose2D Layers

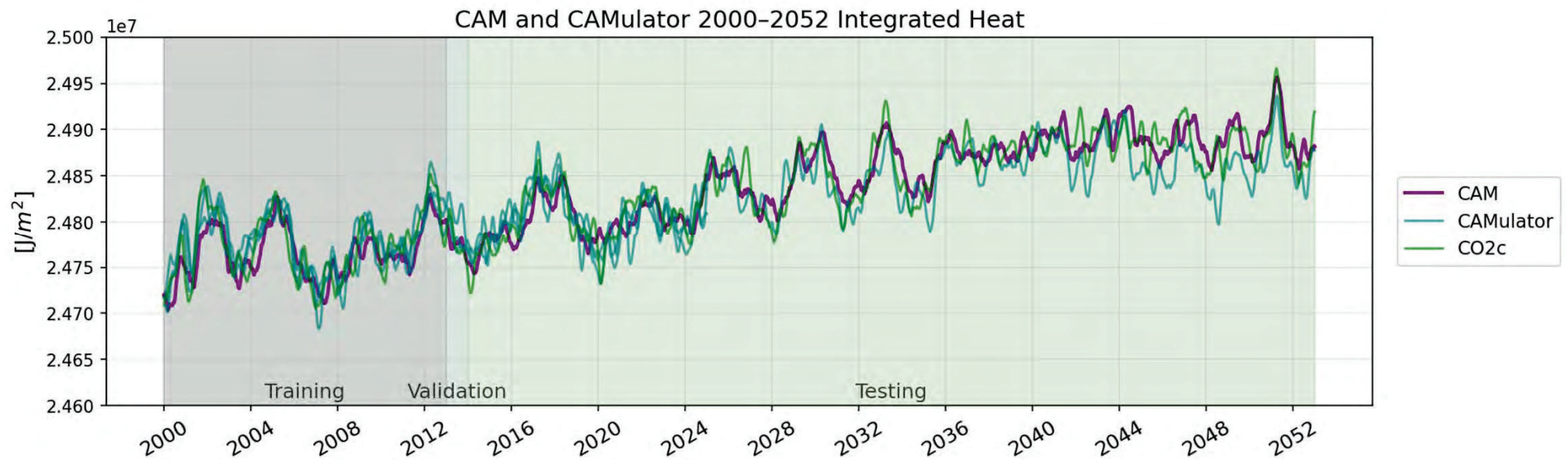
## CAMulator V.1.1



## Constant 2010 SST – “observed” CO2

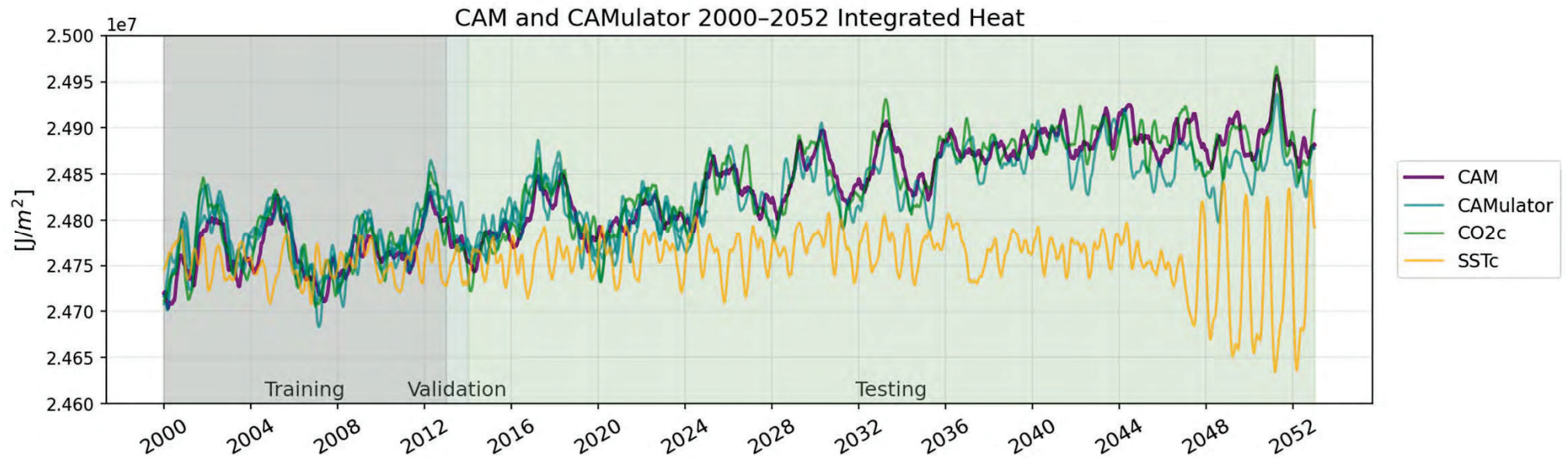


## Constant 2010 CO<sub>2</sub> – “observed” SSTs

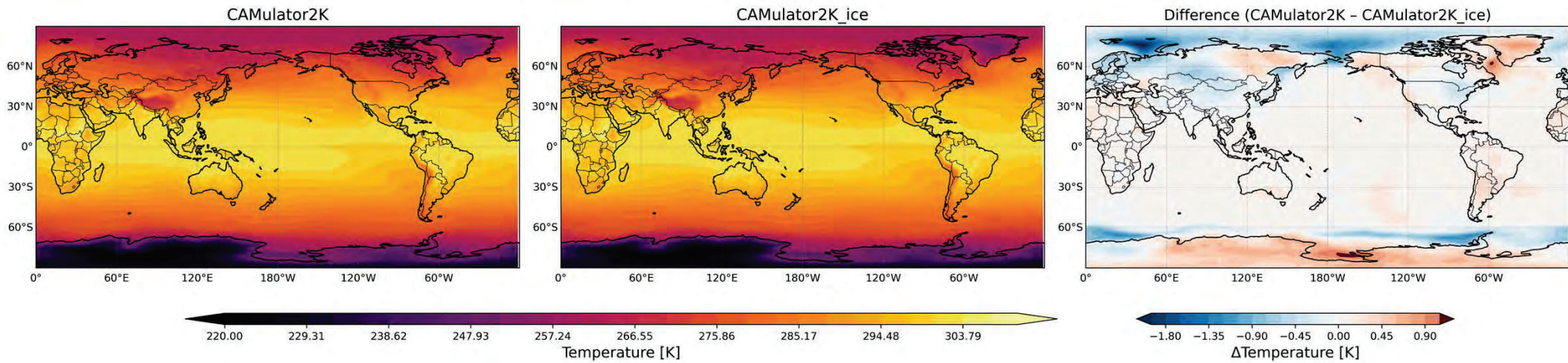




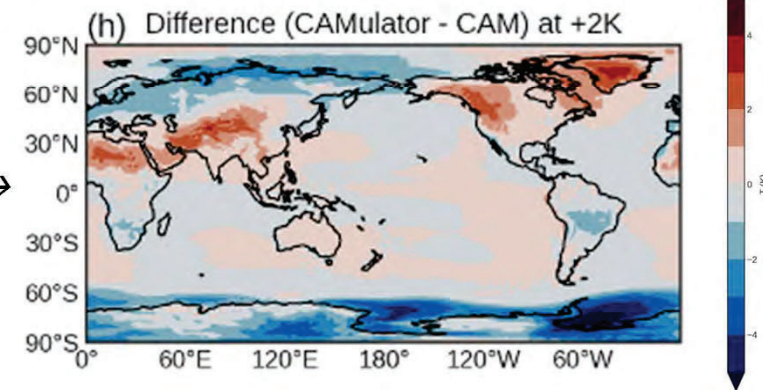
## Constant 2010 SST – “observed” CO2



## Two Meter Temperature Annual Biases: +2K experiment + Sea ICE



**CAMulator V.1.0 Model Biases: →**



## CREDIT Future Directions

### Open Questions

- Ensemble generation: what is the most accurate method with least latency?
- Tradeoffs between data volume, model size, input data size, and types of physical constraints
- How to improve vertical exchange of information in model, especially between troposphere and stratosphere
- End-to-end black box model vs more interpretable/tunable collection of component models?
- Data assimilation: traditional methods versus DA emulators vs hybrid methods

### Next Steps

- Improve usability of CREDIT with software engineering support
- Adding ensemble generation
- Regional model training and evaluation
- S2S and longer scale rollout evaluation
- Coupling to active ocean (ML and otherwise)
- Training a new weather model with more vertical levels at 0.25 degree or finer resolution

## Summary

- CREDIT opens a new pathway to customization of the whole AI weather and climate modeling pipeline
- New paradigm of building AI-ready processes and interacting software
- CREDIT source:  
<https://github.com/NCAR/miles-credit>
- Links to CREDIT papers:  
<https://miles.ucar.edu/projects/credit/>

### Contact Me

Email: [wchapman@ucar.edu](mailto:wchapman@ucar.edu)

Github: willychap

Version 2025.2.0 is out now!



Q Search

#### Getting Started

Getting Started

Installing CREDIT from source

#### Configuration File

What's in the Configuration File?

#### Training and Inference

Training a Model

Running Inference

Evaluation and Metrics

#### Contributing

Contributing

#### Adding New Models and Datasets



## MILES-CREDIT Documentation

Welcome to the documentation for **MILES-CREDIT**, the **NSF NCAR Community Research Earth Digital Intelligent Twin** project. CREDIT is a machine learning-based research platform for understanding the best practices for training and operating global and regional AI autoregressive models, built as part of the NSF NCAR **Machine Integration and Learning for Earth Systems (MILES)** group.

CREDIT enables users to train, run, and evaluate AI-based numerical weather and climate models. This documentation will guide you through installation, configuration, training, inference, evaluation, and extending the system with custom datasets and models.

#### What you'll find here:

- How to install CREDIT from source
- How to set up and train a model
- How to run inference and evaluate results
- How to contribute datasets, models, and enhancements
- Config file reference for reproducible HPC runs
- Tutorial videos for visual guidance

If you encounter issues or have suggestions, please open an issue on our GitHub repository. Contributions are welcome!

#### Getting Started

[Getting Started](#)

[Installation for Single Server/Node Deployment](#)

[Installation on Derecho](#)