



ESMO

Earth System Modelling
and Observations



Machine Learning in Earth System Modelling and Observations

Douglas Rao (NC State University)

Working Group on Observations for Researching Climate (WGORC)

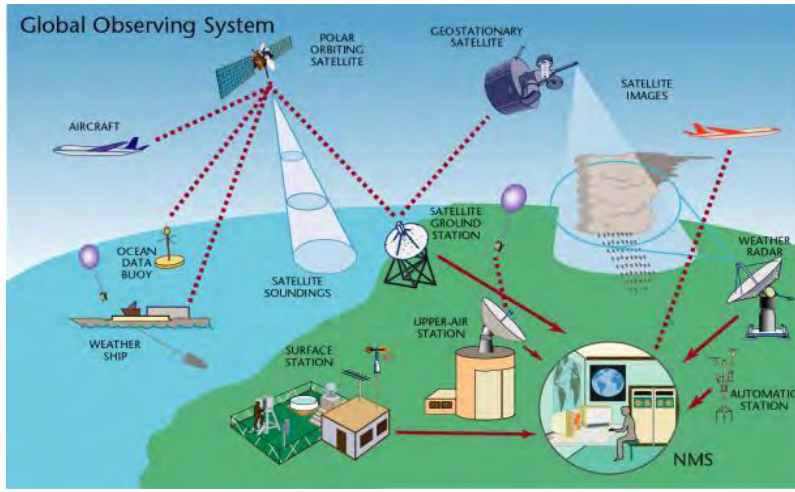
WMO Workshop on Needs and Application of Climate Data Management to
Support the State of the Climate Reporting

23–26 June 2025, Astana, Kazakhstan

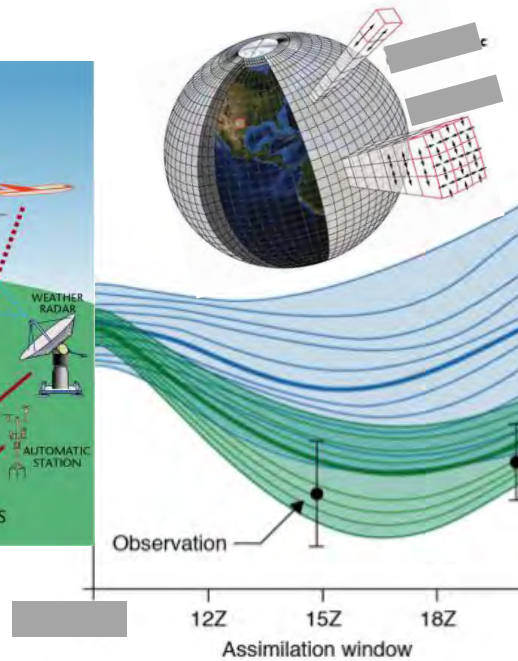
What is ESMO?

- A new WCRP Core project that coordinates and advances all WCRP **observations, data assimilation** and **modelling**
- Works on three objectives:
 - **Advancing predictions and projections of the Earth system** on time scales from weeks to centuries and furthering model-observation integrated frameworks.
 - **Improve monitoring, understanding, and attribution of Earth system changes and impacts** with robust uncertainty quantification through the synthetic use of models and observations.
 - **Advancing and harnessing emerging technologies** in modelling and observations.

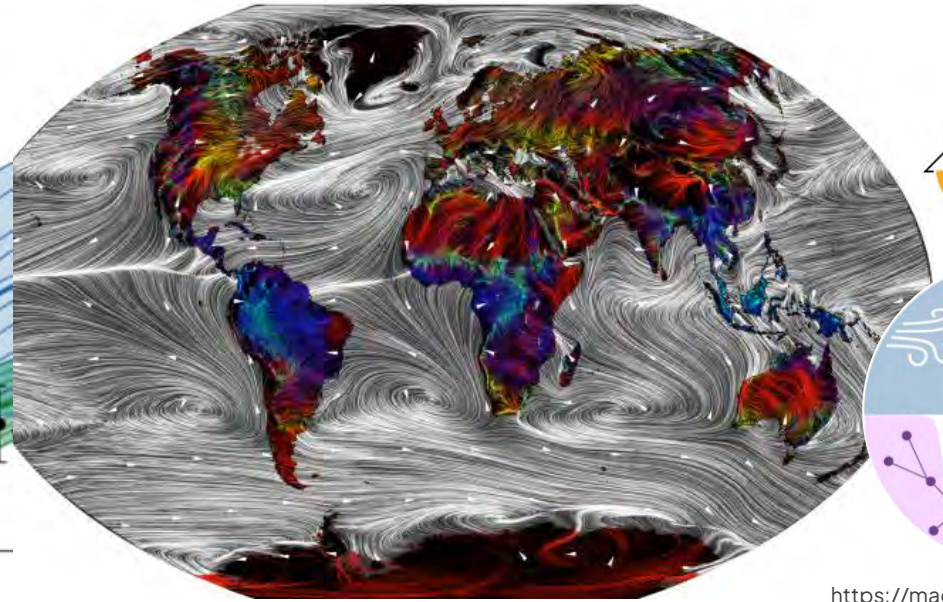
What is ESMO?



WMO Global Observing system



Cambridge University Press Climate Model and ECMWF Data Assimilation schematic



Kling and Ackerly, 2020



<https://machineclimate.de/>

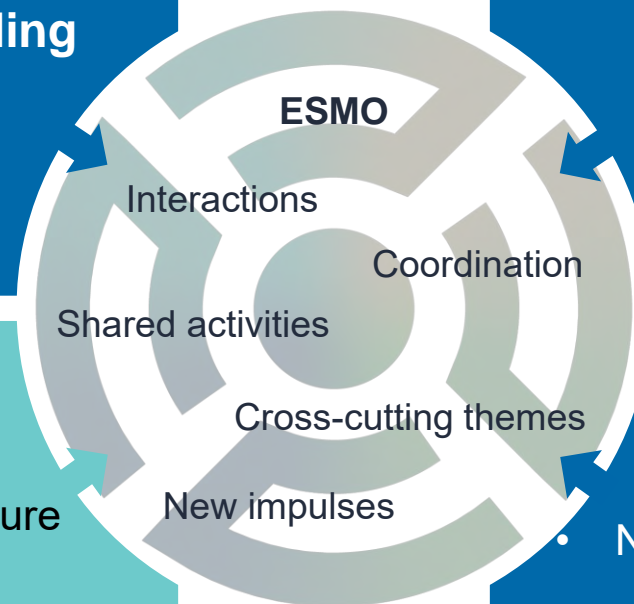
Modelling Community in WCRP

Working Group on Coupled Modelling (WGCM)

- Development and review of coupled climate models

Coupled Model Intercomparison Project (CMIP)

- Understanding of past, present and future climate changes
- Assessment of model performance



Working Group on Numerical Experimentation (WGNE)

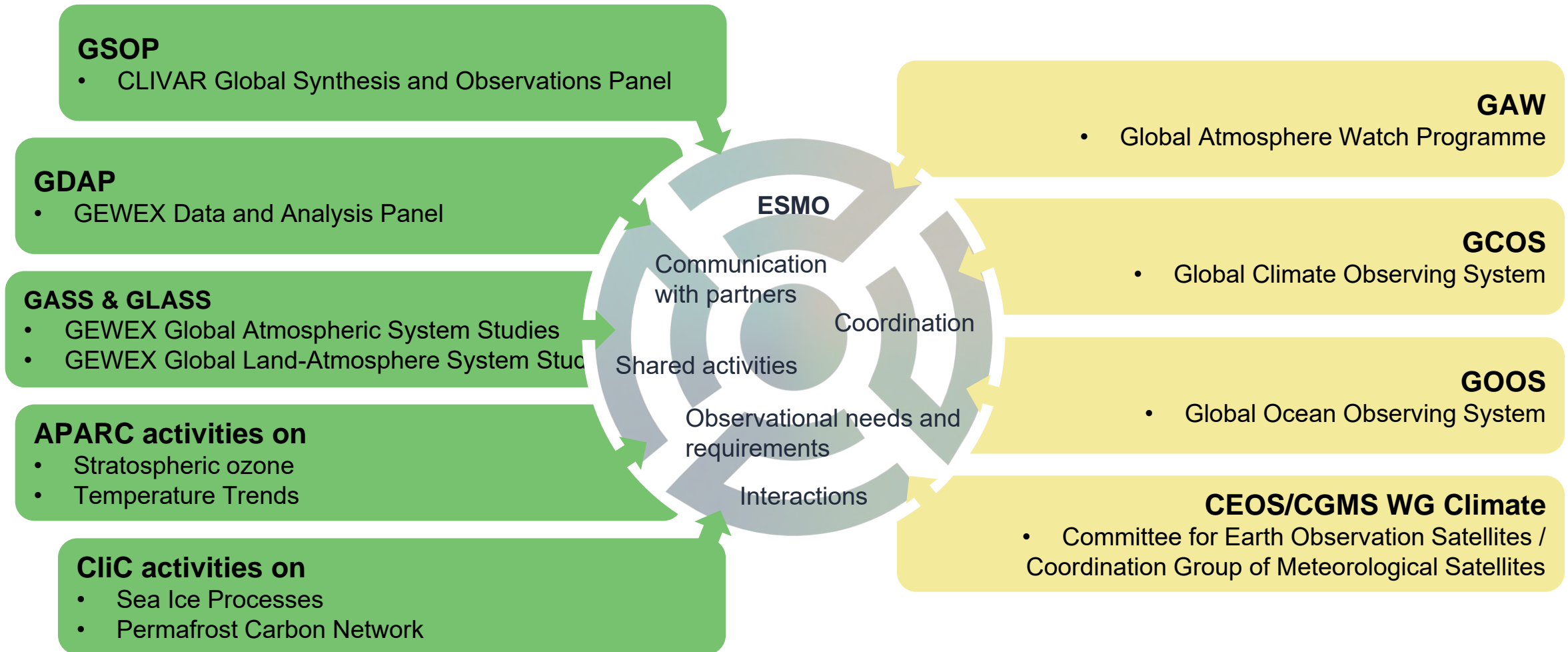
- Development of Earth system models (design, implementation, error diagnosis, revisions)

Working Group on Subseasonal to Interdecadal Prediction (WGSIP)

- Numerical Experimentation for S2I variability and predictability

Observational Community in WCRP

Topic-specific observational groups in core projects



ESMO Objectives



Advancing predictions and projections of the Earth system



Improving monitoring, understanding and attribution of climate system changes and impacts



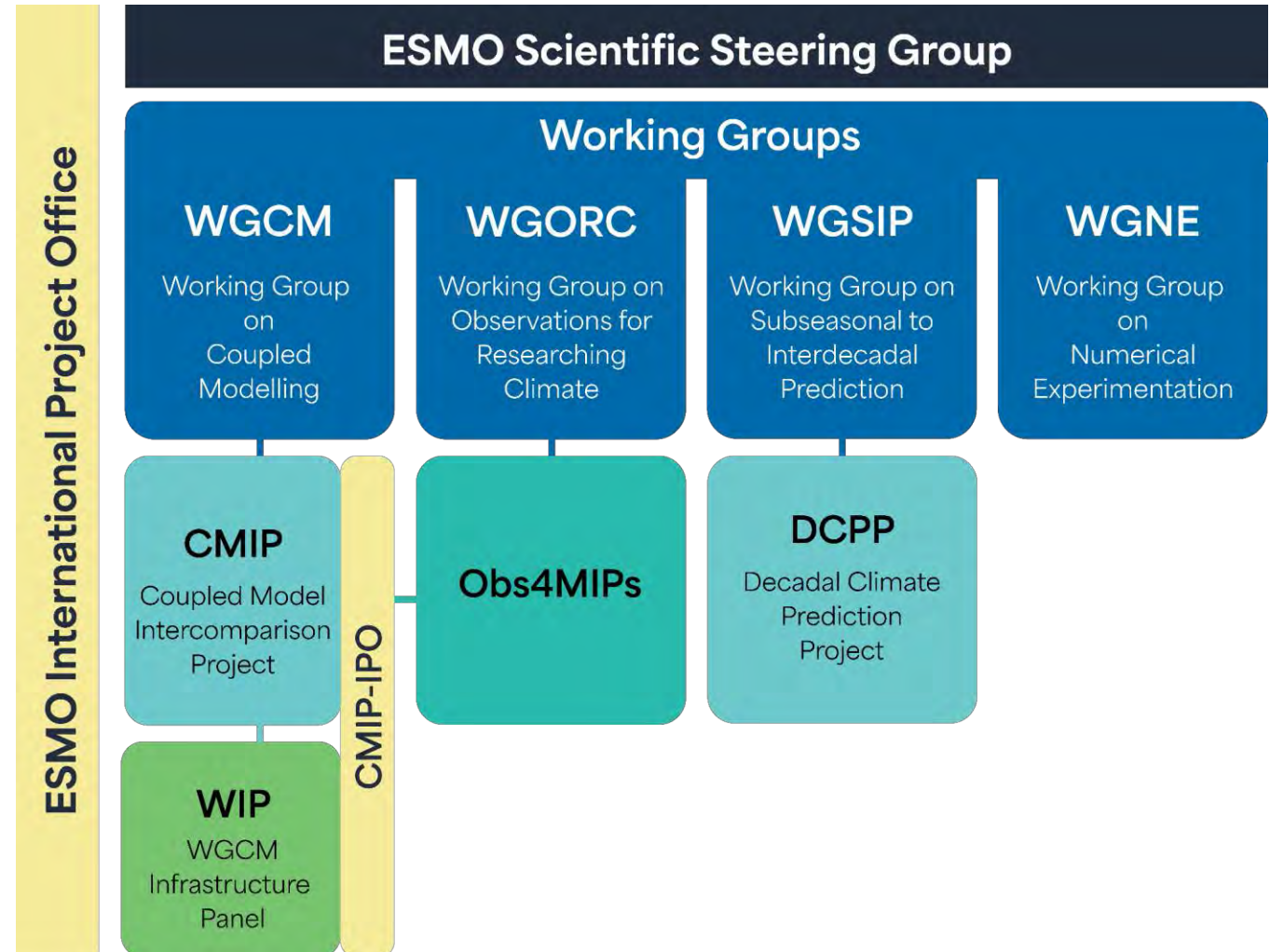
Advancing and harnessing emerging technologies

Requires an **integrated** and **consistent framework** combining global Earth system observations, data assimilation and modelling.

These objectives will be addressed in collaboration with the WCRP core projects, Lighthouse Activities and Working Groups.

Working Group on Observations for Researching Climate (WGORC)

- Co-chairs:
 - Douglas Rao (NC State University)
 - Amy Doherty (Met Office)
- ESMO SSG Liaison:
 - Claire MacIntosh (ESA ECSAT)
 - Alison Cobb (ECMWF)
- Coordinate research and development at the intersection of observation and modeling interface.



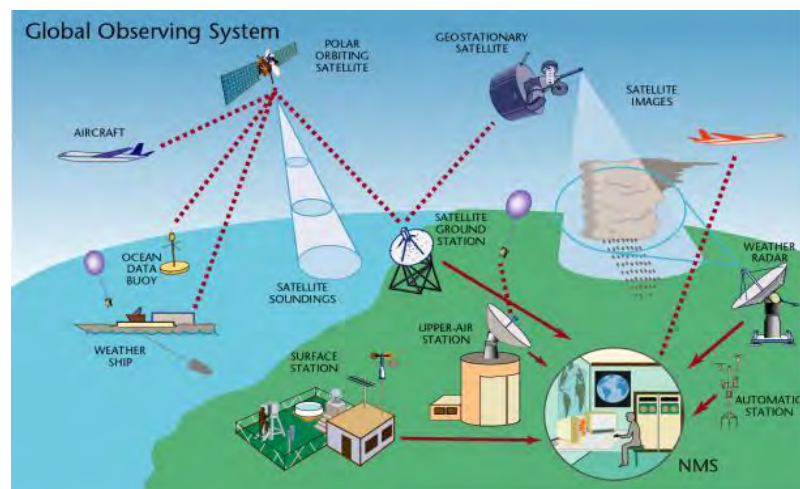
Machine Learning in Earth System Modeling & Observation

Quality Control

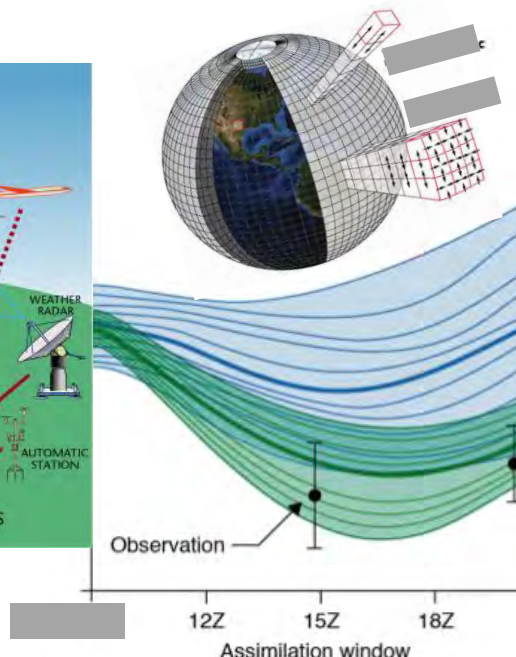
Data fusion

Assimilation

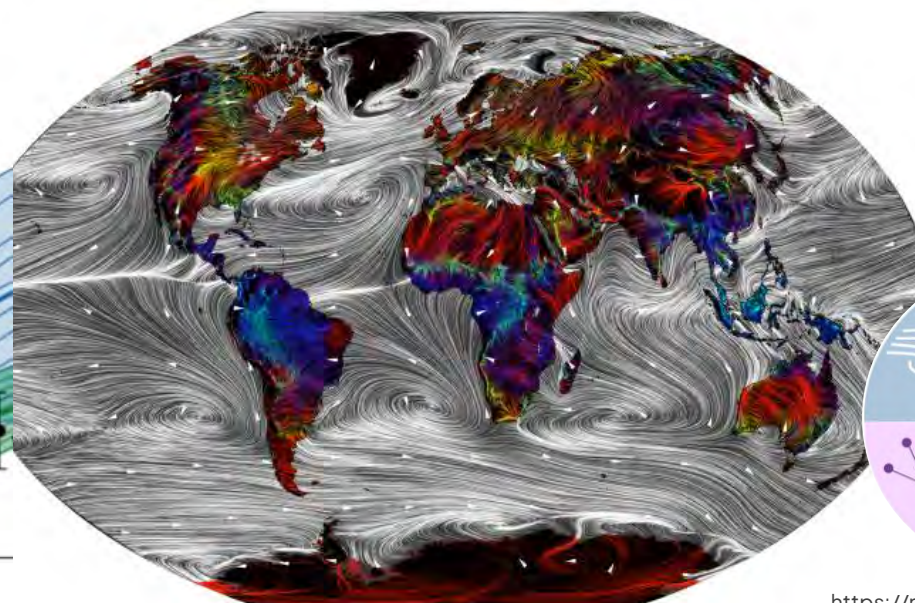
Emulation



WMO Global Observing system



Cambridge University Press Climate Model and ECMWF Data Assimilation schematic



Kling and Ackery, 2020



<https://machineclimate.de/>

Reconstruction

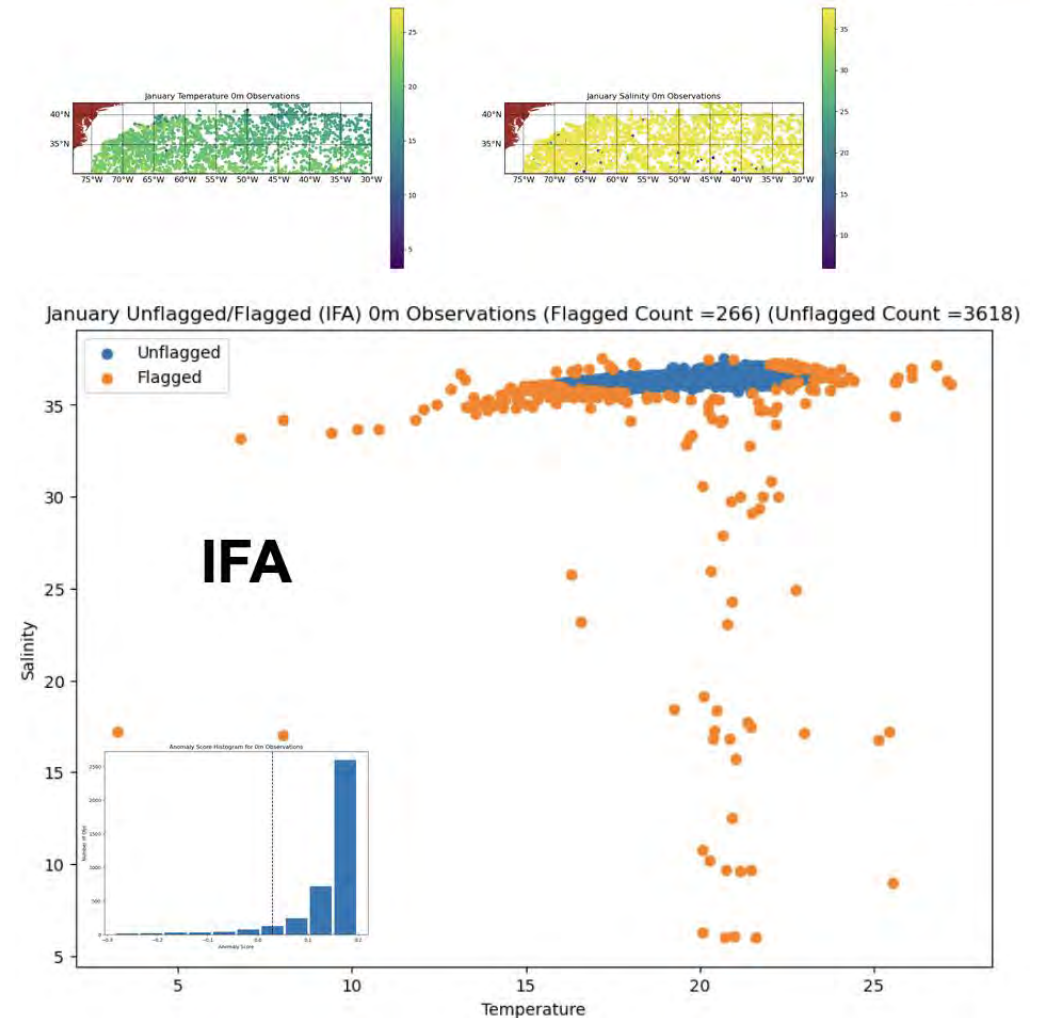
Parameterization

Data-driven simulation

ML for Data Quality Assurance

Source: Jim Reagan

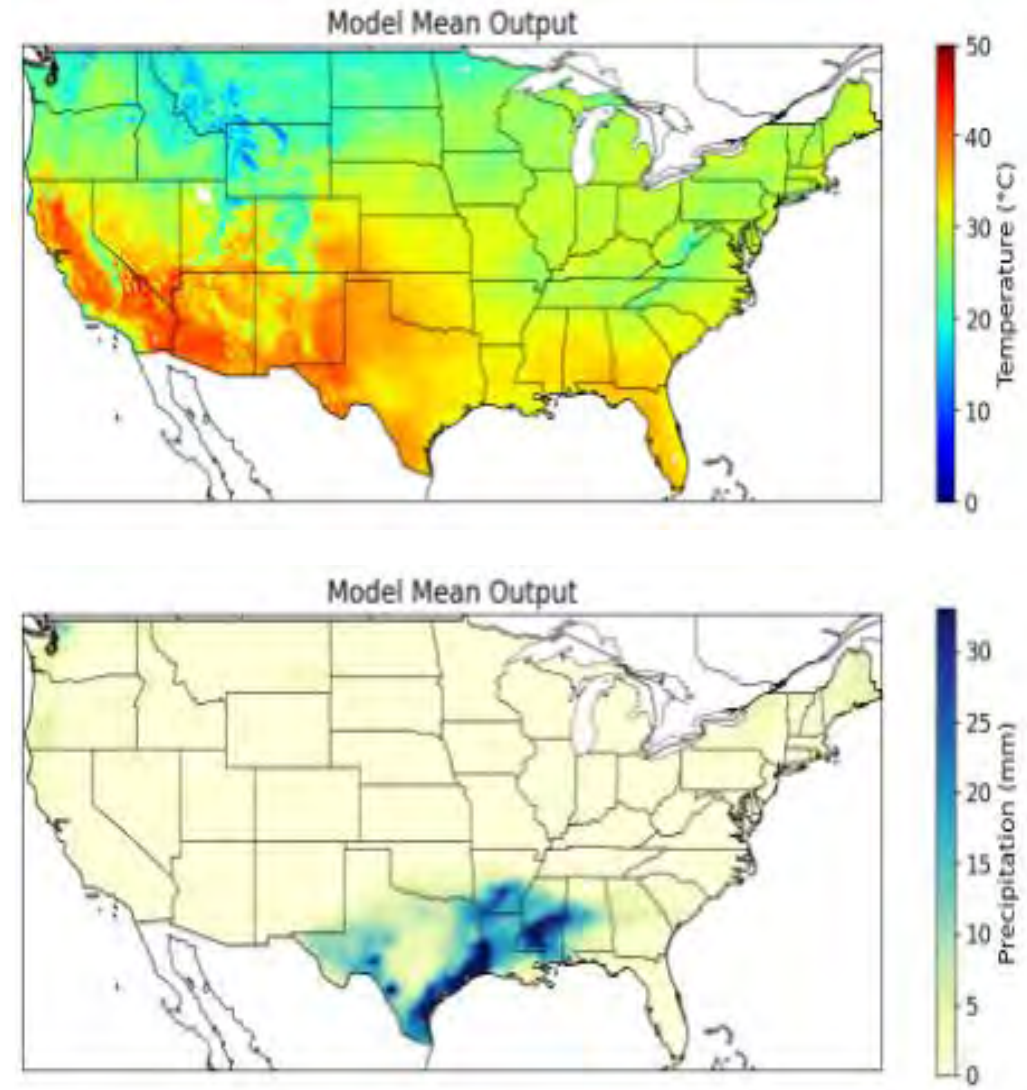
- Machine learning is skillful in learning from data patterns and identifying anomalous data in time series.
- Common use of ML for QA
 - Anomaly detection
 - Change point detection
 - Bias correction
- Example pilot use cases:
 - US Climate Reference Network
 - Ocean profiles data



ML for Data Reconstruction

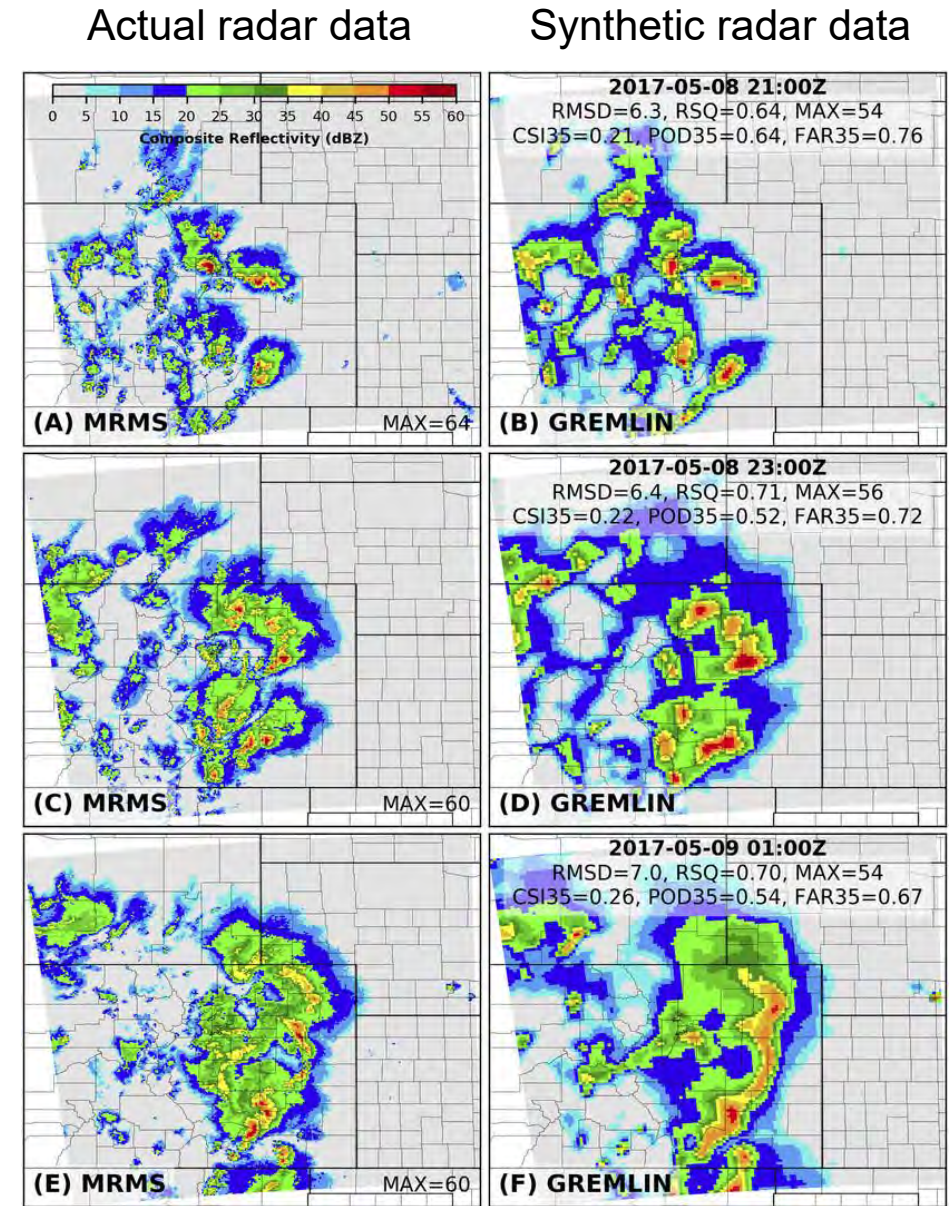
Source: Douglas Rao

- ML's ability to represent nonlinear relationship in data could benefit climate data reconstruction.
- Example use cases:
 - NOAA GlobalTemp (v6.0):
<https://doi.org/10.25921/rzxg-p717>
 - ERSST (v6.0):
<https://doi.org/10.1175/JCLI-D-23-0707.1>
 - High-resolution gridded climate dataset (right figure)

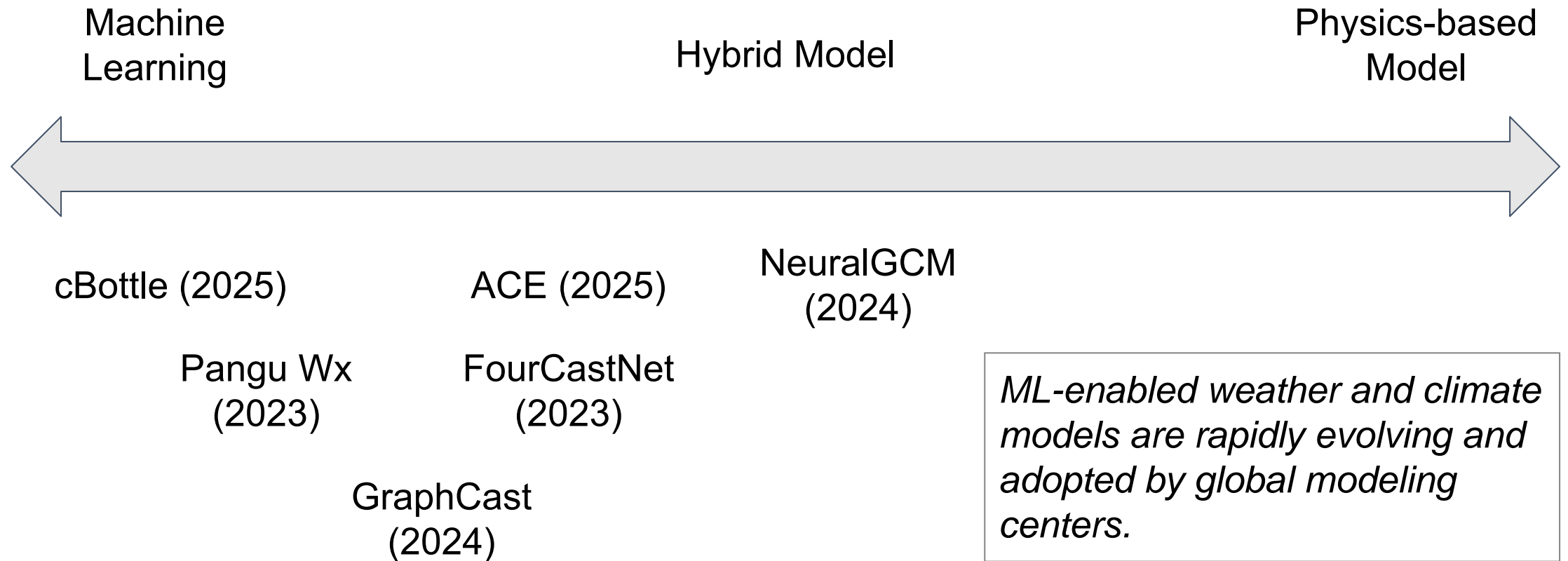


ML for Climate Data Fusion

- Machine learning can be used to bring different observations (e.g., *in situ* & satellite) together for better climate monitoring.
- Increasingly, ML has been used to generate synthetic data where traditional observations are unavailable.

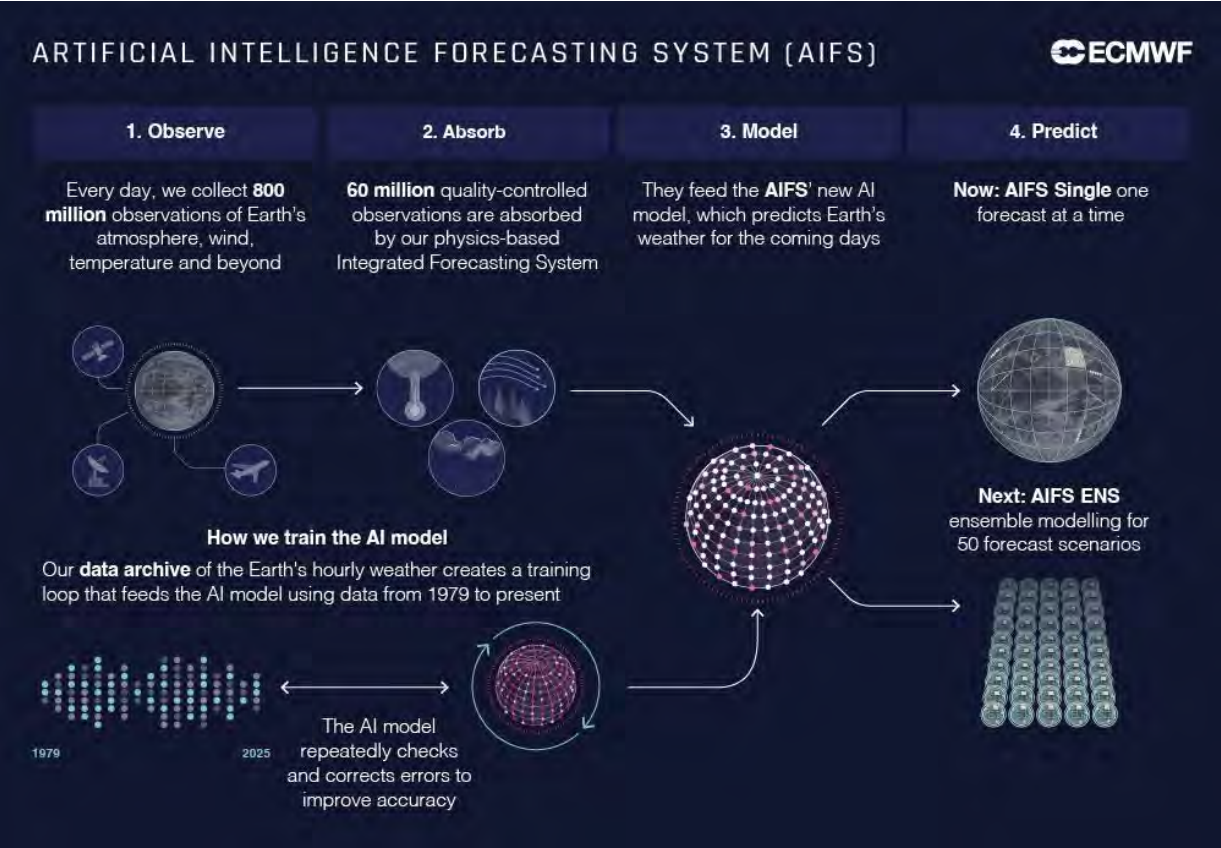


Continuum of Physics-based & ML-based modeling



ECMWF's AI forecasts become operational

25 February 2025

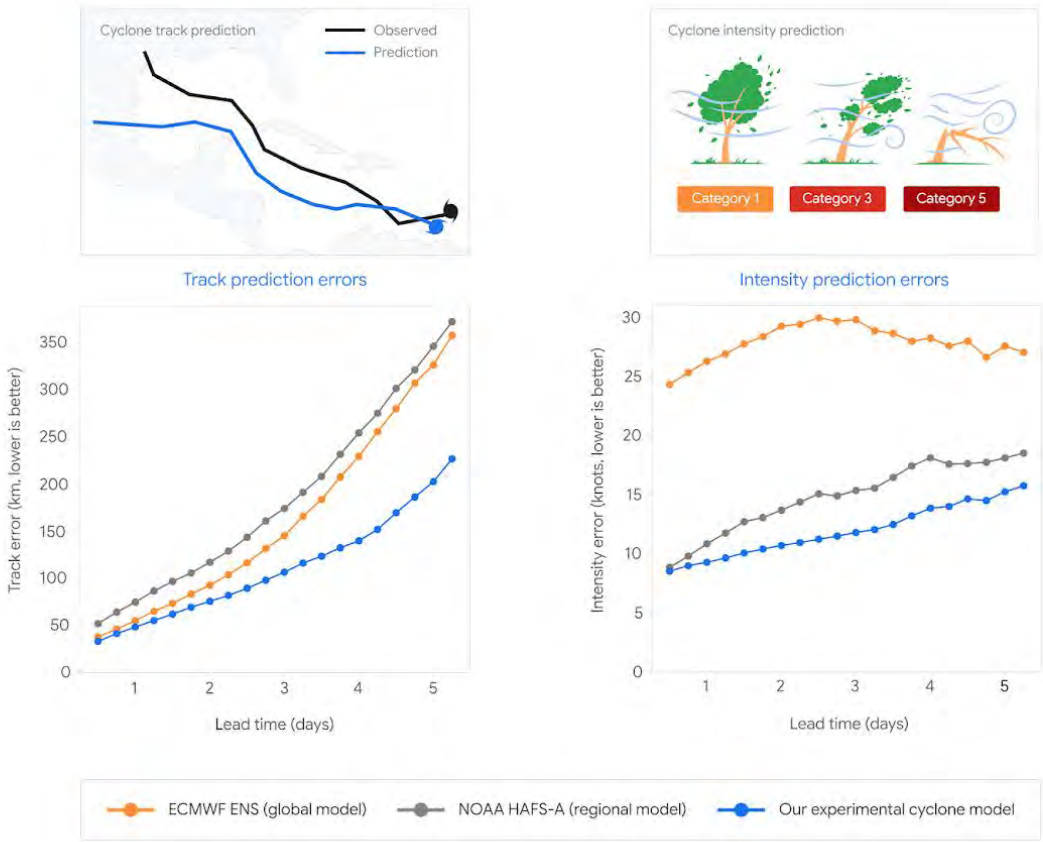


Source: ECMWF

How we're supporting better tropical cyclone prediction with AI

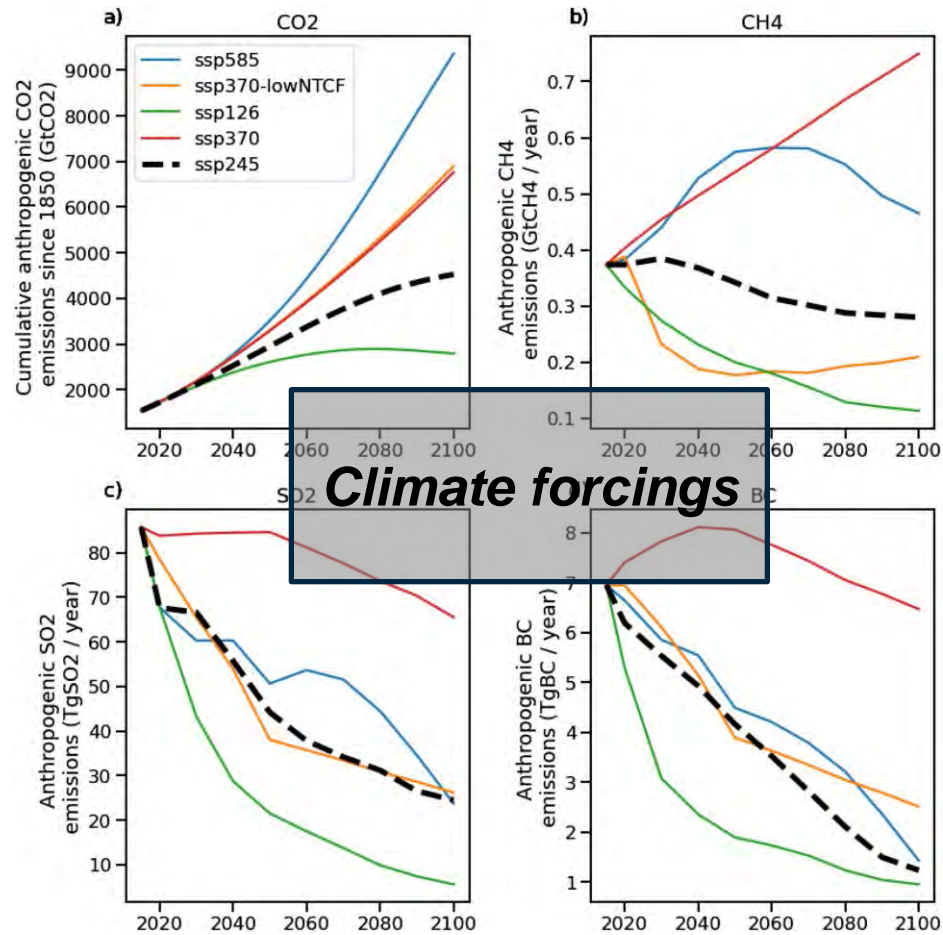
RESEARCH

12 JUNE 2025

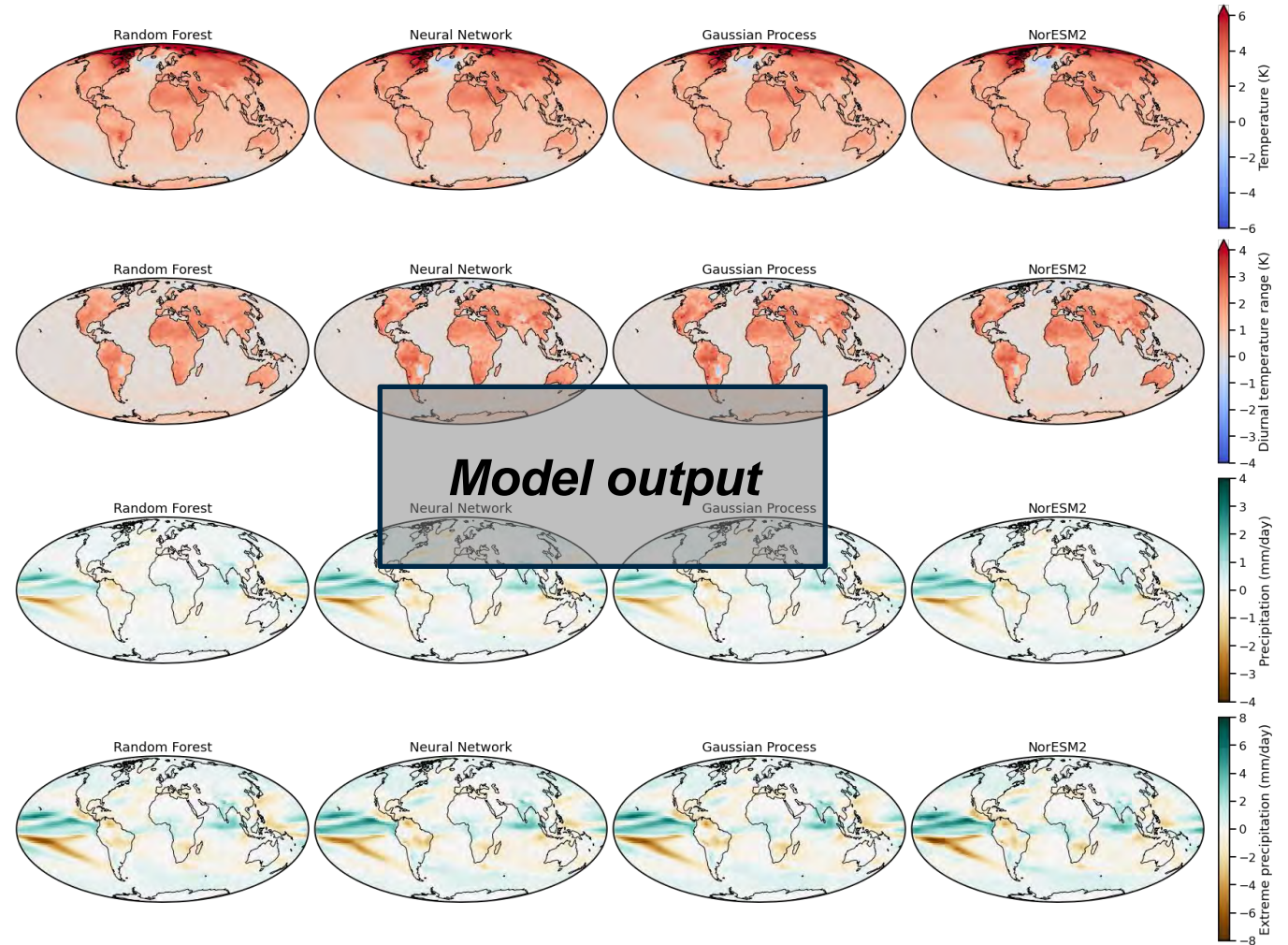


Source: Google Research

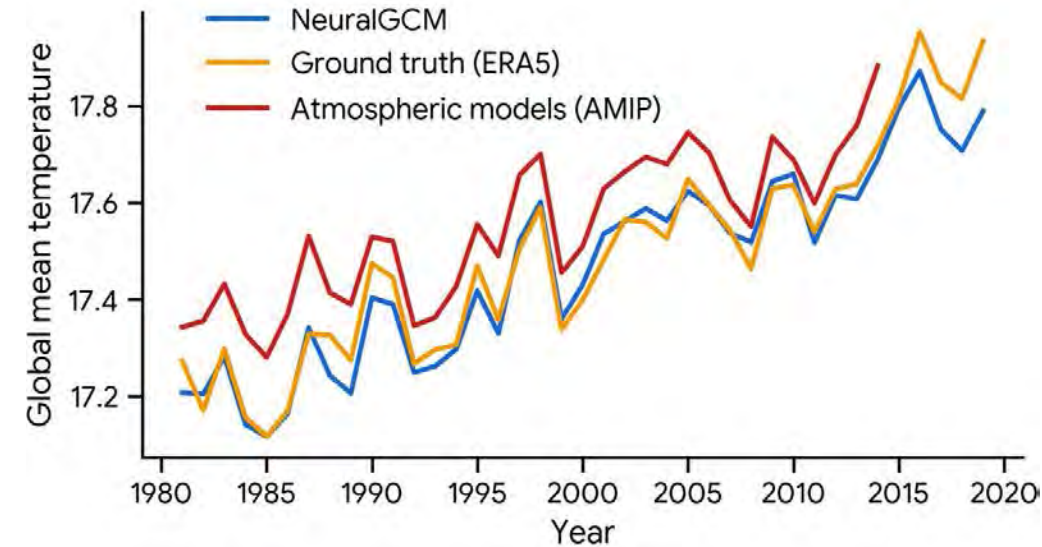
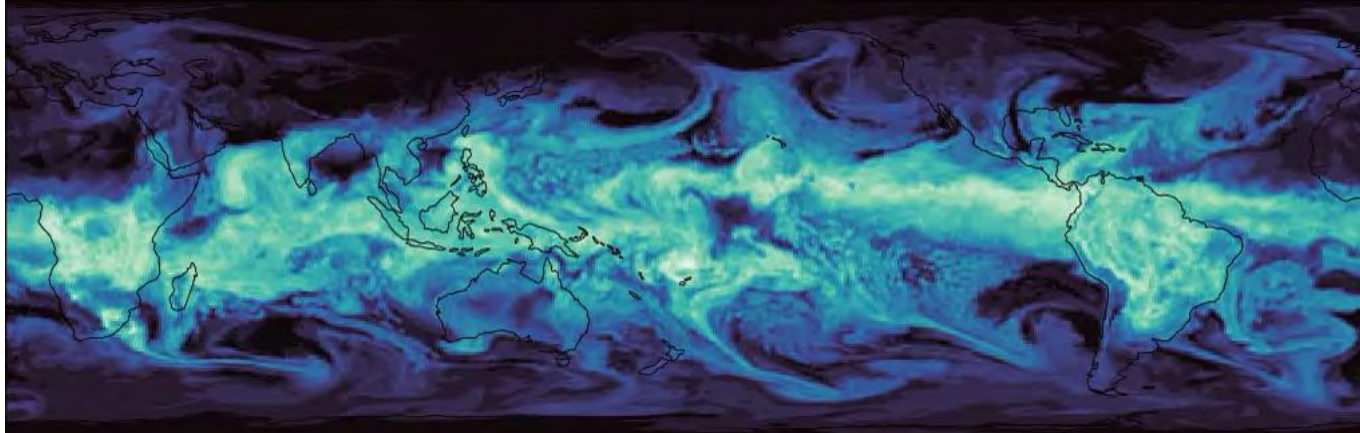
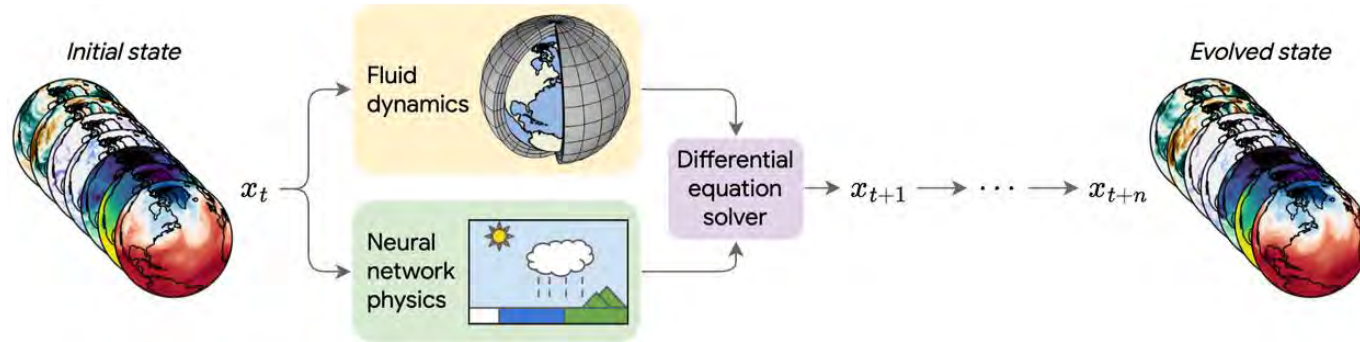
Climate Model Emulation



Climate forcings



Data-driven & Hybrid Modeling



Source: Stephen Hoyer (2024)
<https://doi.org/10.1038/s41586-024-07744-y>

Data-driven & Hybrid Modeling

Model Run Time



Time Simulated By Model

NeuralGCM



0.0 days

NCAR CAM6



0.0 days

NOAA X-SHIELD



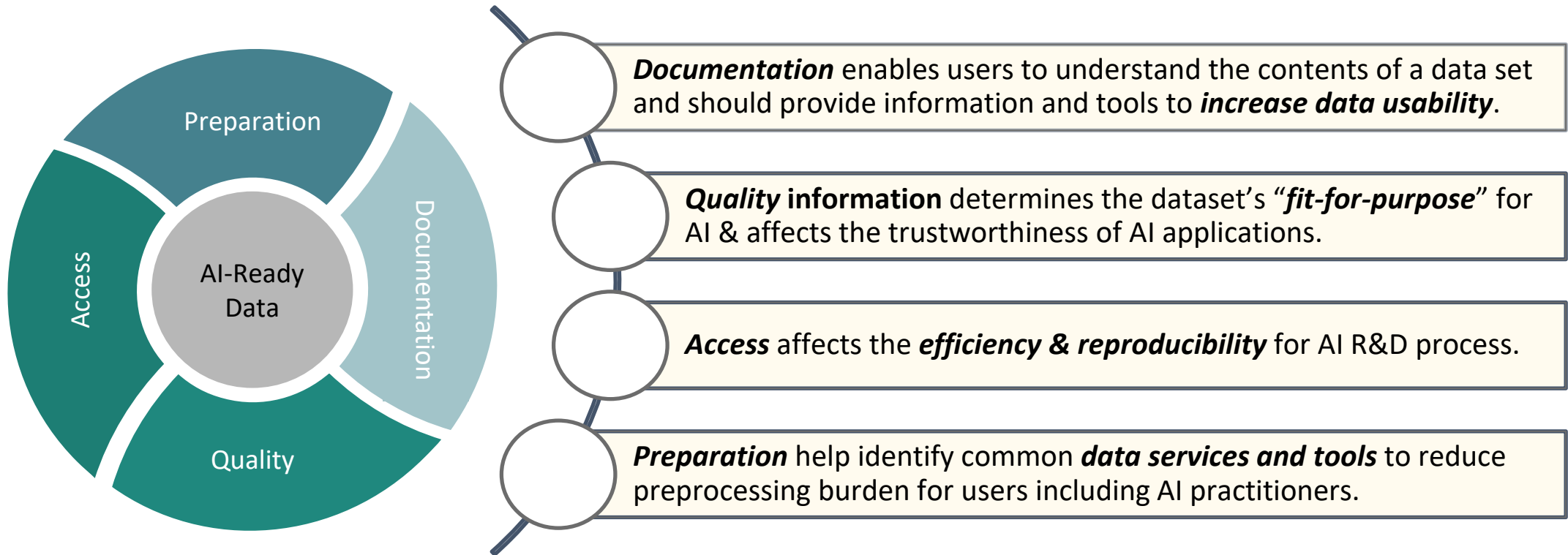
0.0 days

Source: Google Research
<https://research.google/blog/fast-accurate-climate-modeling-with-neuralgcm/>

Issues to Consider for Data Stewardship

- ML could and should be leveraged to improve the climate data for various applications.
- **Governance** of the ML-based climate data is challenged by the increasing volume, diversity, and producers.
- **Reproducibility** of ML-based climate data is still an issue.
- **Uncertainty** quantification of ML-based climate data is less mature.
- ML-based climate data is fundamentally limited by existing climate datasets because of the model training workflow.

Modernizing Data Management for AI/ML



Earth Science Information Partners AI-ready data checklist (<https://github.com/ESIPFed/data-readiness>) is developed as a framework to assess the readiness level of open environmental data for AI applications.



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Thank you!

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