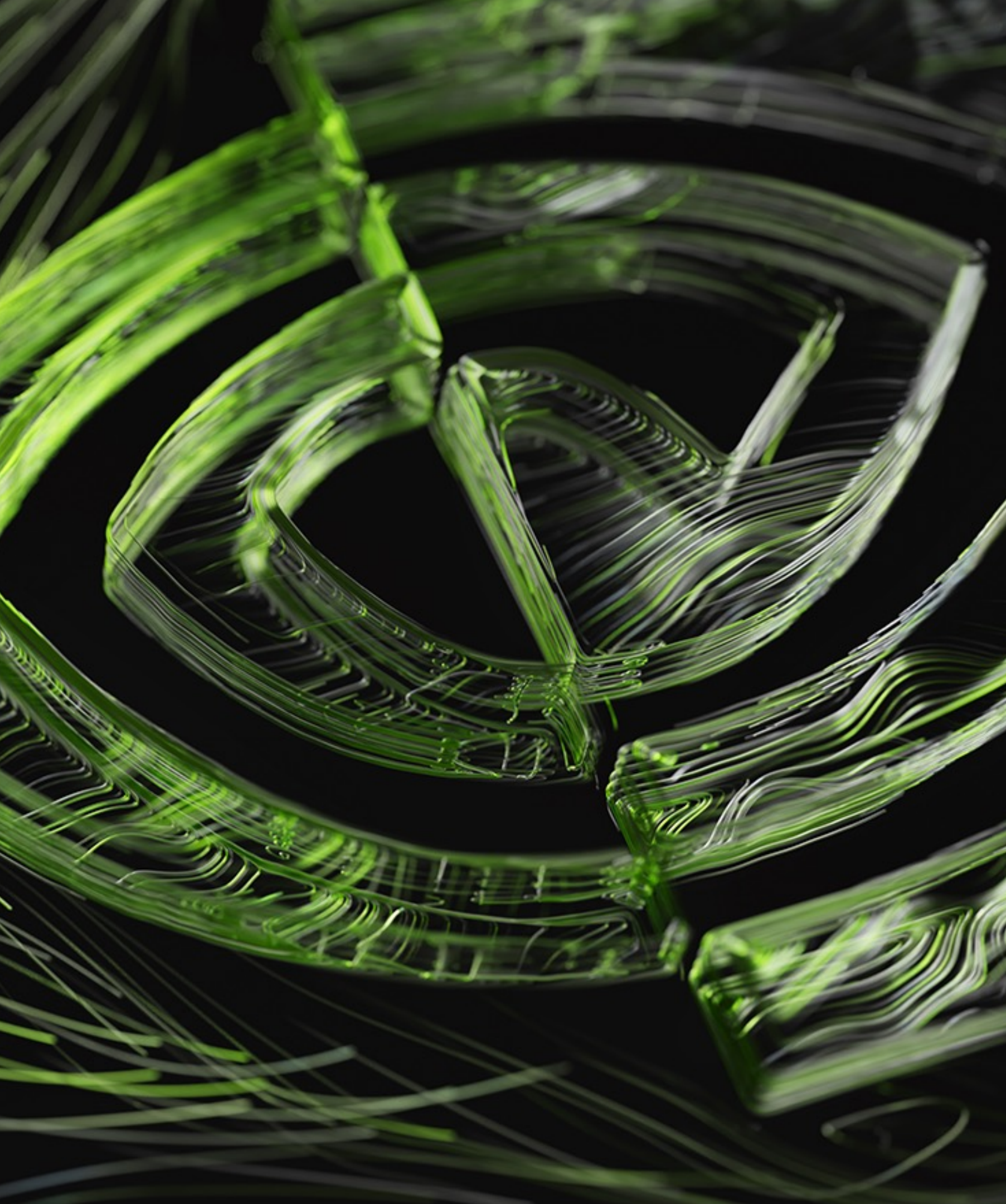




# Earth-2: Digital Twins for Weather and Climate

Karthik Kashinath, Principal Engineer and Scientist, AI-HPC

Engineering Lead, NVIDIA Earth-2 Initiative



# Agenda

- NVIDIA's Earth-2 initiative: The Big Picture

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- FourCastNet & latest accomplishments in Weather.

---
- Beyond Weather, towards Climate.

---
- Digital Twinning platforms

---

# The Future Under Climate Change will be Harsh

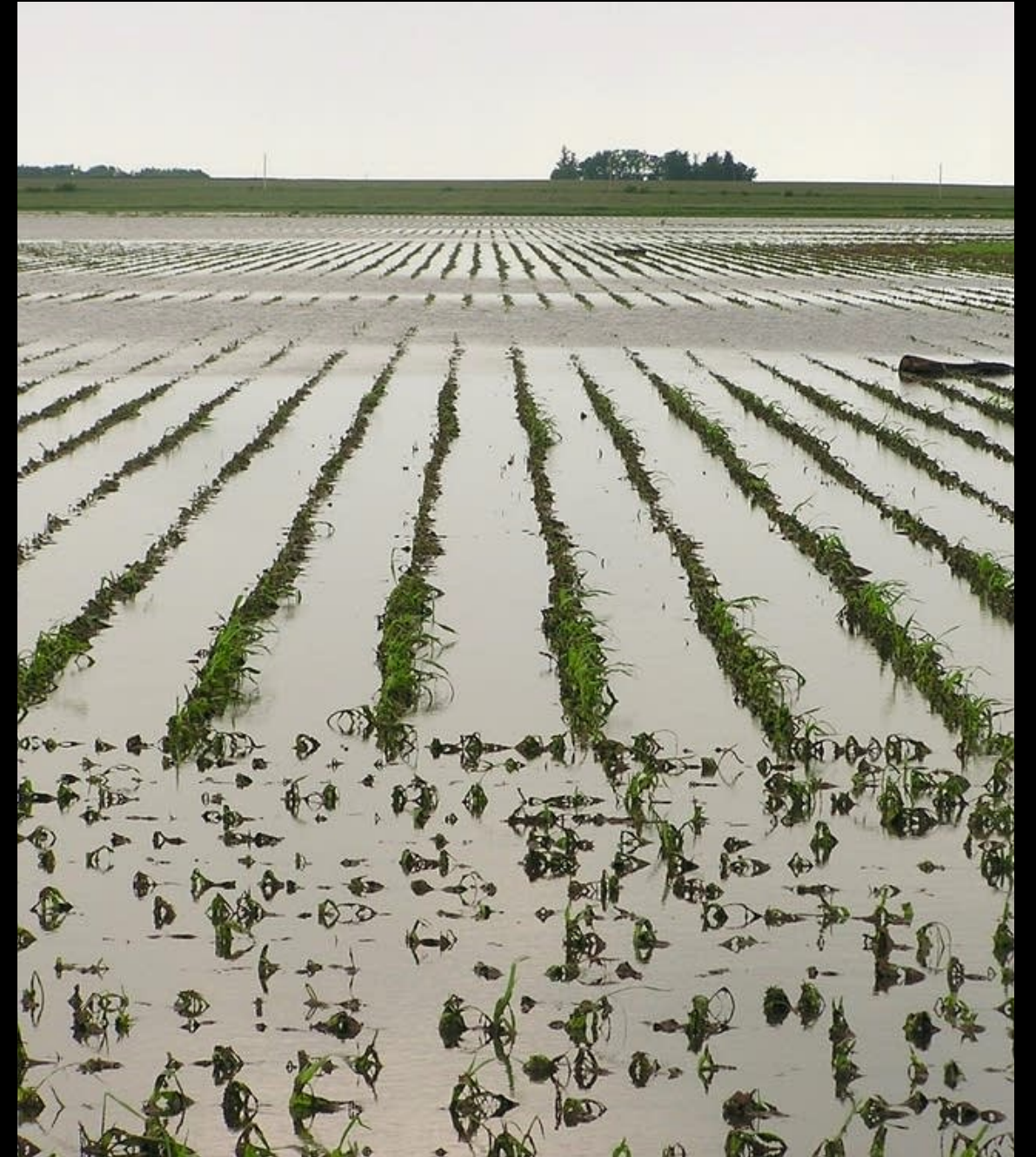
We urgently need better tools to prepare for it



WILDFIRE PREVENTION



WATER MANAGEMENT



CROP FORECASTING

# High-Resolution Climate Prediction is a Computational Challenge

Today's climate models are too low resolution. Brute force numerical solvers are decades away from what is needed.

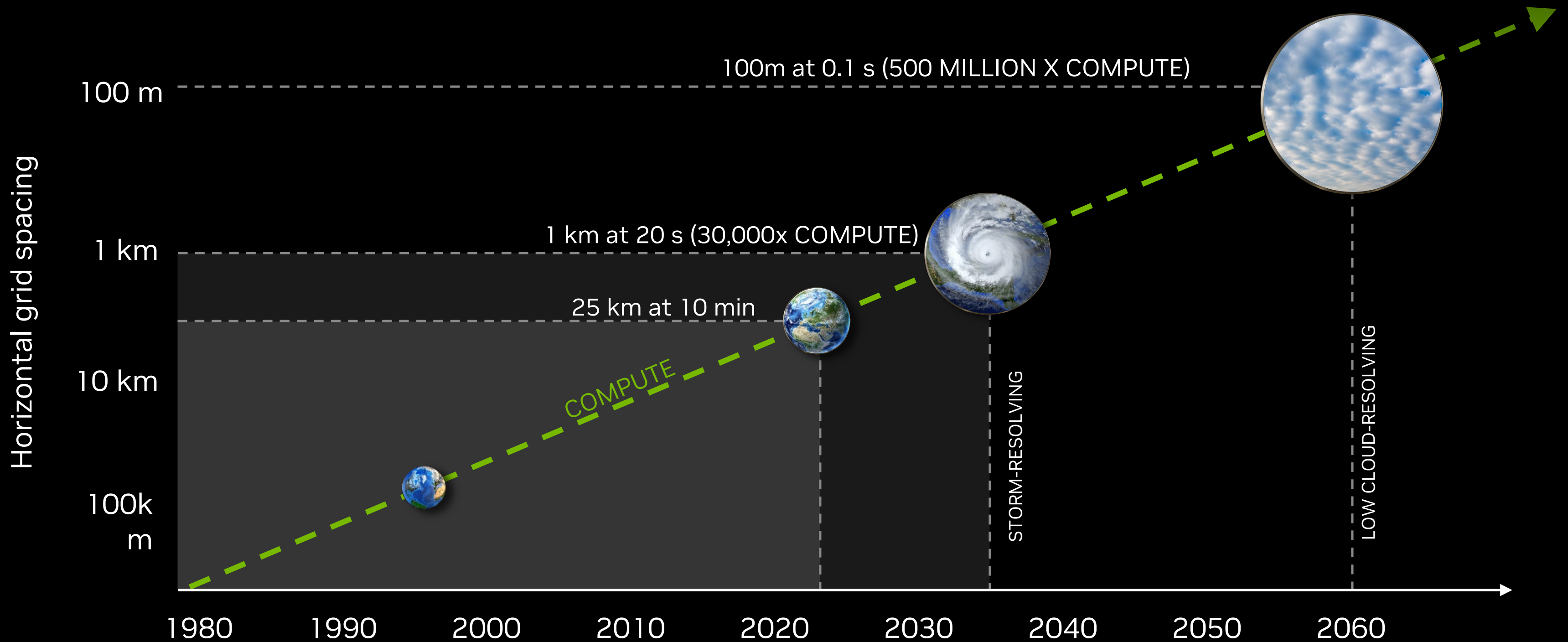
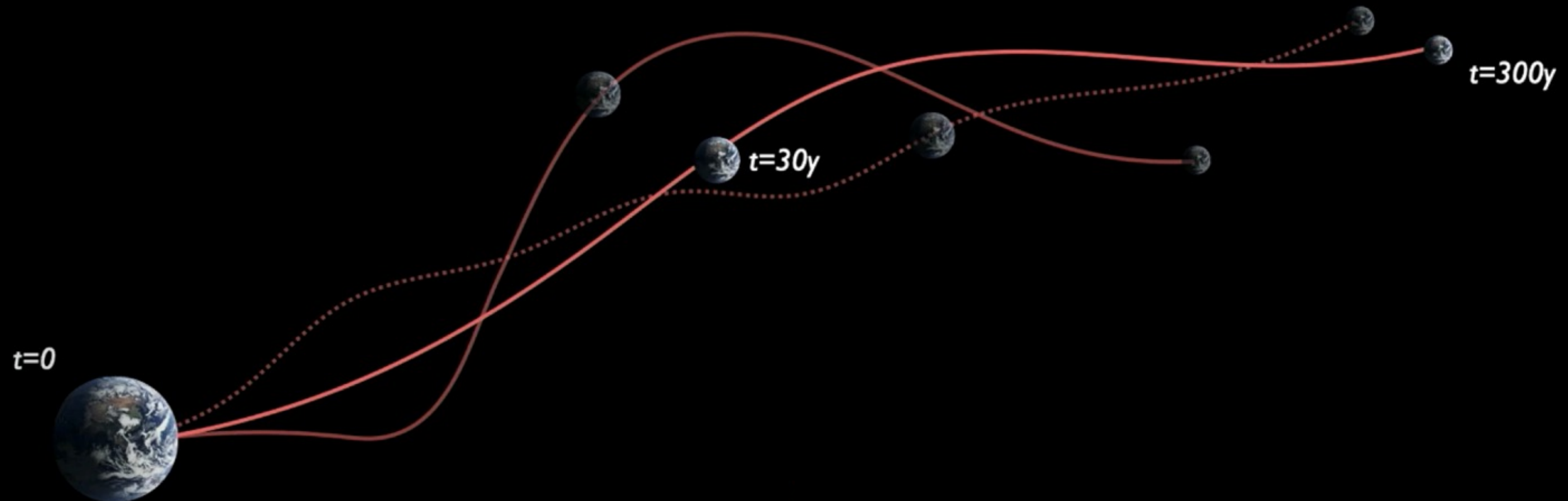


Figure adapted from: Schneider, T., Teixeira, J., Bretherton, C. et al. "Climate goals and computing the future of clouds". *Nature Climate Change* 7, 3-5 (2017)

# It is Hard to Interact with High-Resolution Climate Predictions

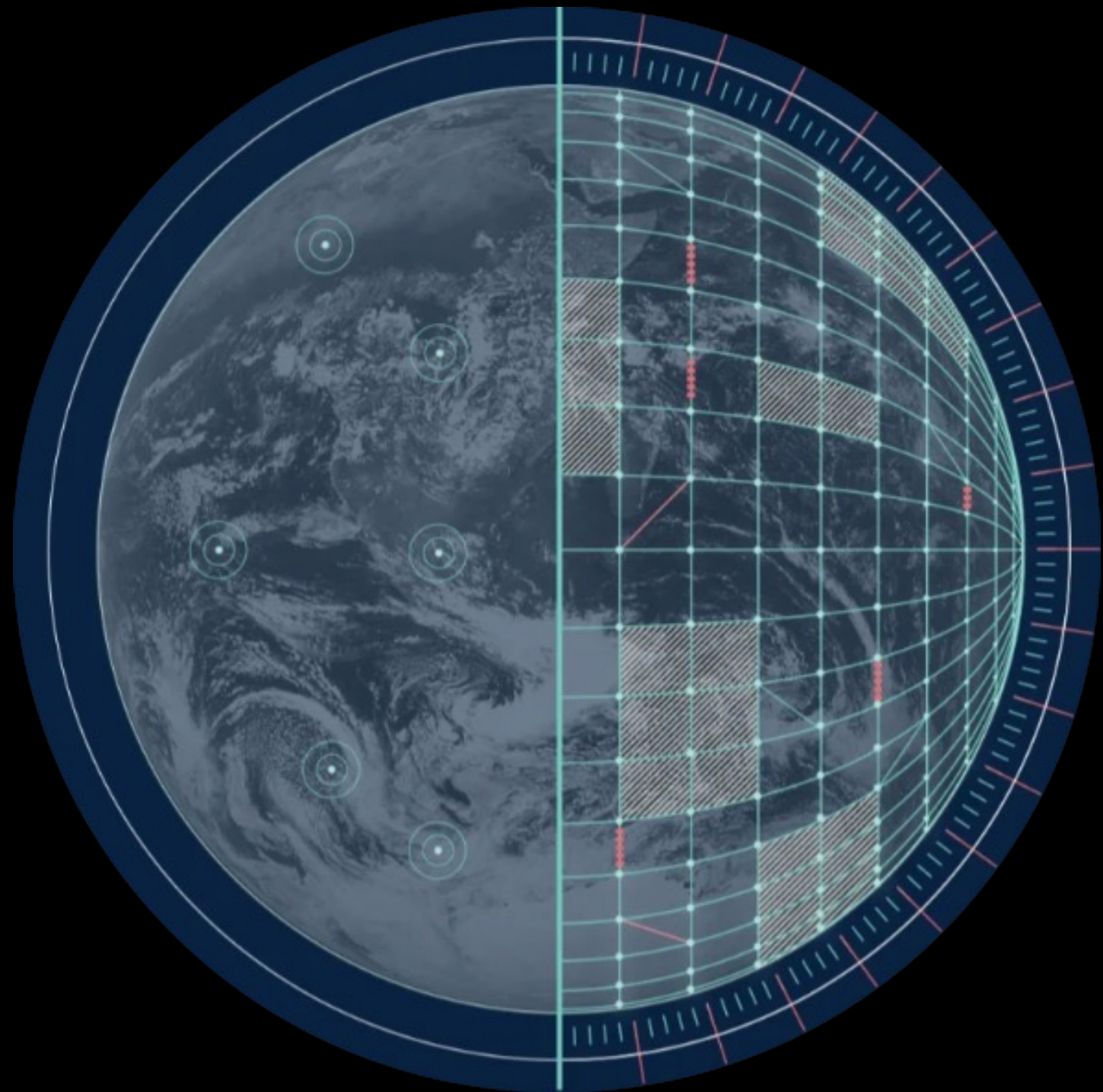
"We can compute km-scale predictions, but can't effectively extract information content, let alone interact with it"

-- Prof. Dr. Bjorn Stevens.



# EARTH-2: Began as a vision

Of a highly interactive climate information system for serving society with next-gen climate predictions.



Imagine you could Select a Region of the Planet...



# ... Ask Questions about Climate Change's Impacts...

On Food, Health, Infrastructure, Energy systems, and more...





# ... and Receive Useful, Visual & Statistical Guidance?

From a Highly Interactive Future Climate Information System, at High Resolution, that Serves Society...

Earth

Twin Earth

## Earth-2 Mission #1

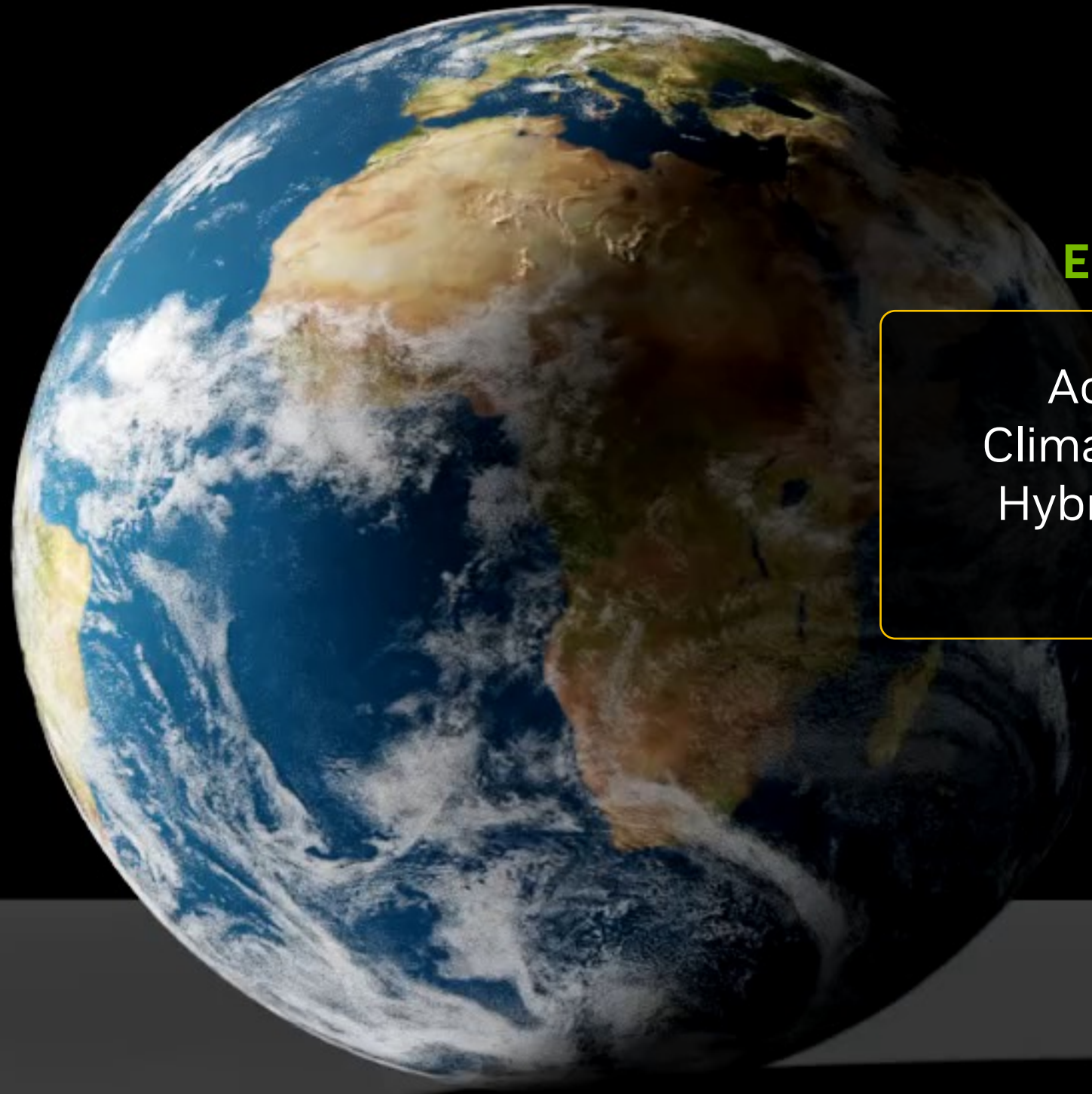
Interacting with Climate  
Predictions at Low Latency.



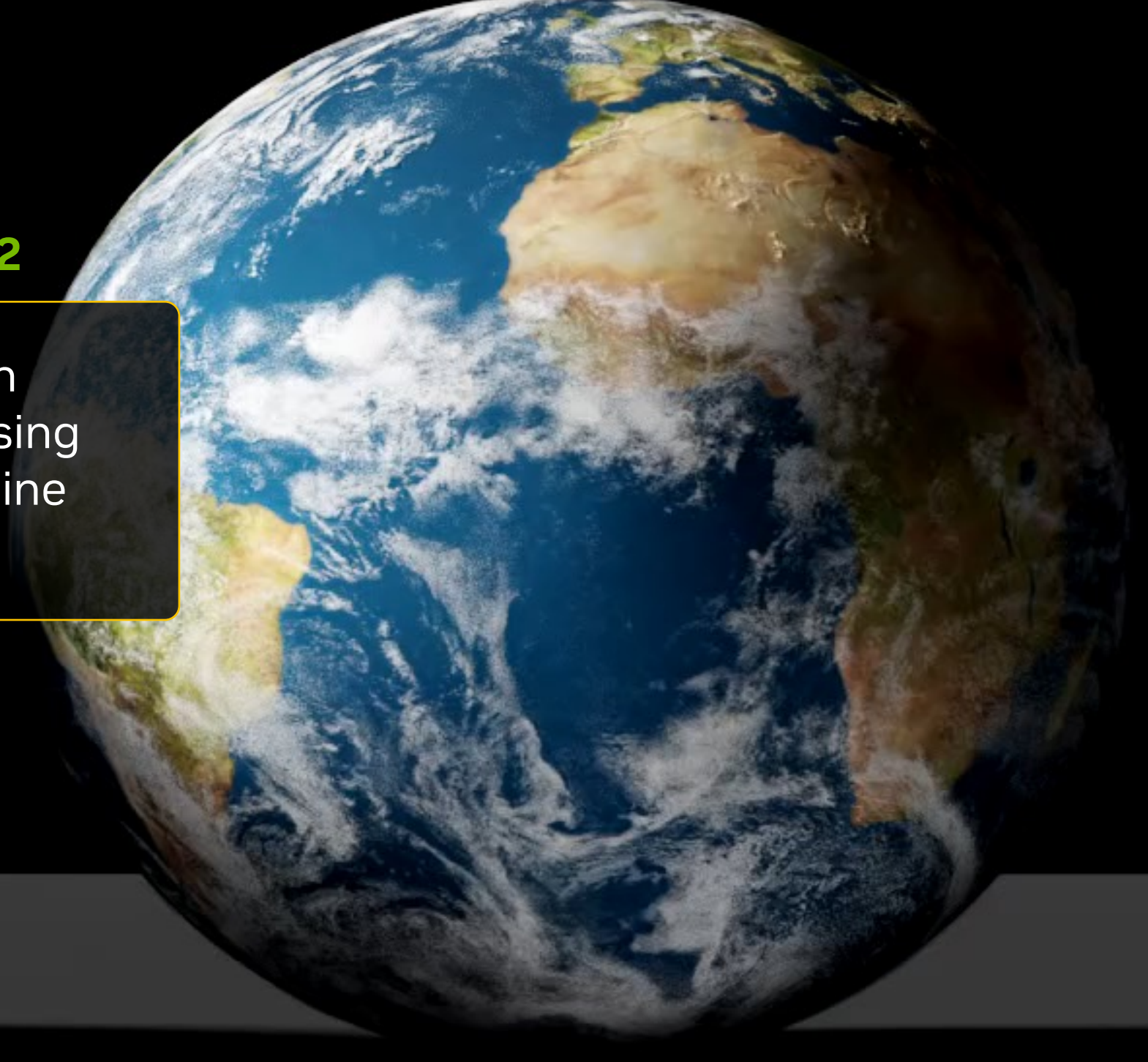
# Imagine the System Evolving in Scientific Fidelity and Computational Ambition

Eventually fed by a new library of climate predictions so high-resolution they seem impossible today.

Earth



Twin Earth



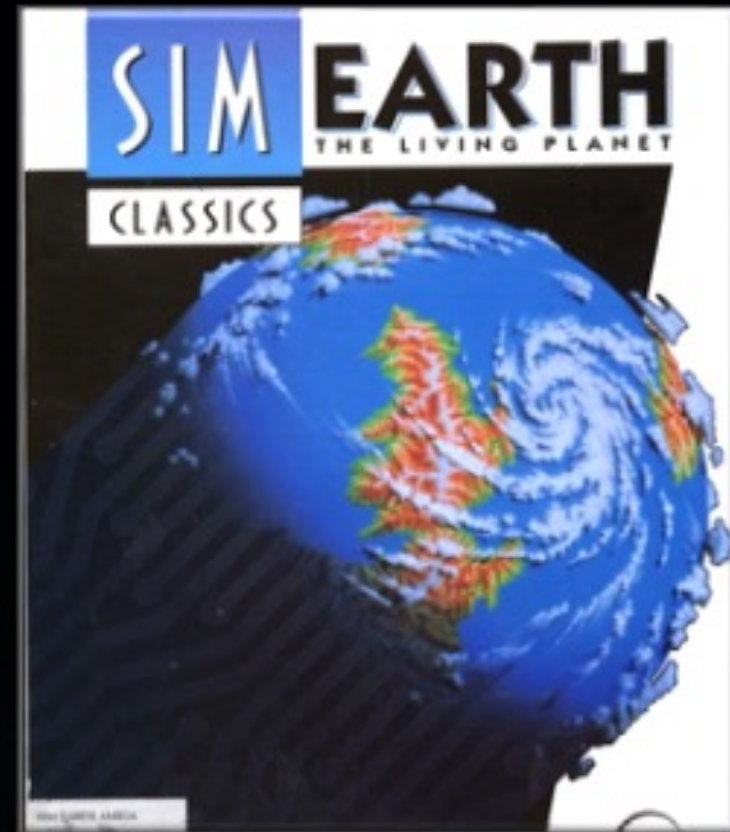
## Earth-2 Mission #2

Achieving Next-Gen  
Climate Predictions using  
Hybrid Physics, Machine  
Learning & HPC.

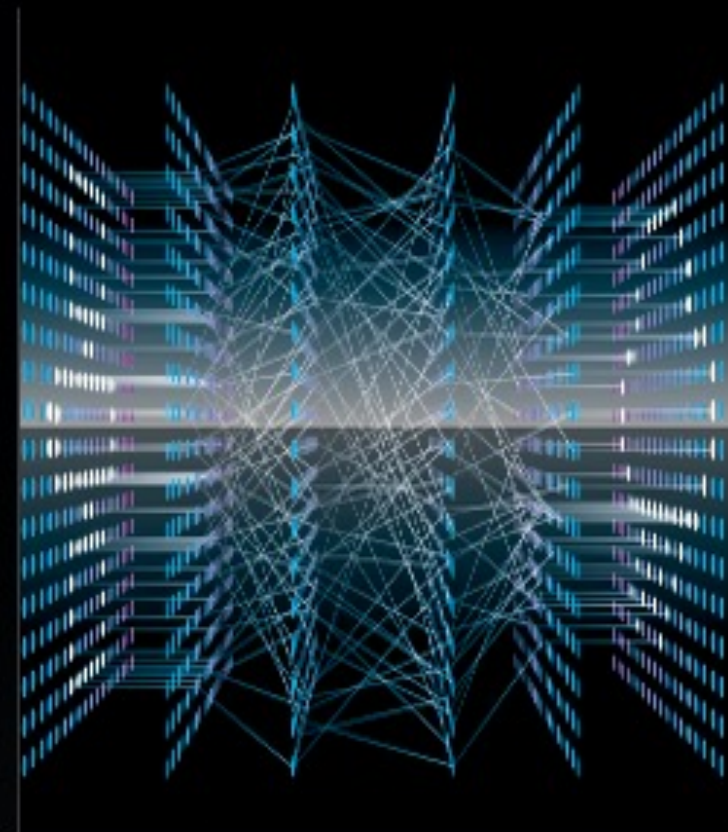
# The World is Already Working Hard on these Problems

Example: Project Destination Earth envisions what Digital Twins could be.

Interactive Collaborative Platform



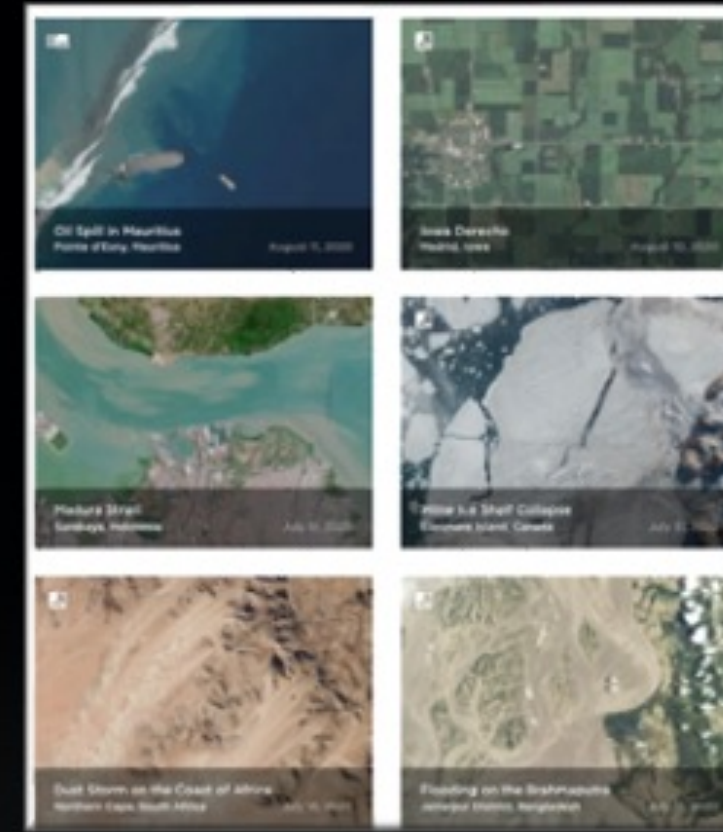
Data-driven Models



Storm-resolving Models



Unified Observations



Exascale Compute

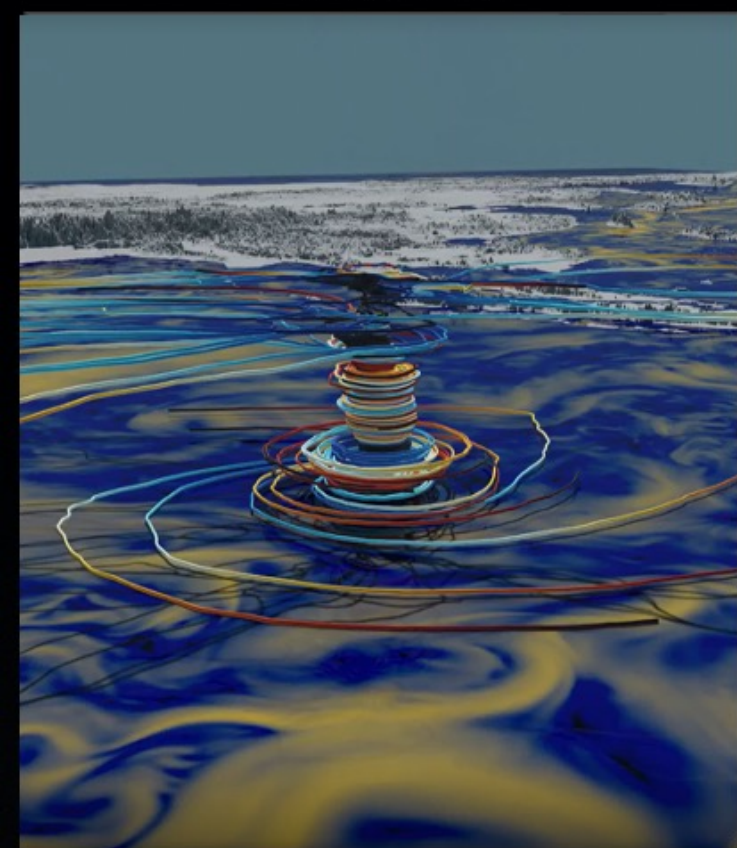


<https://digital-strategy.ec.europa.eu/en/library/destination-earth>

# NVIDIA's technical know-how can make a big difference

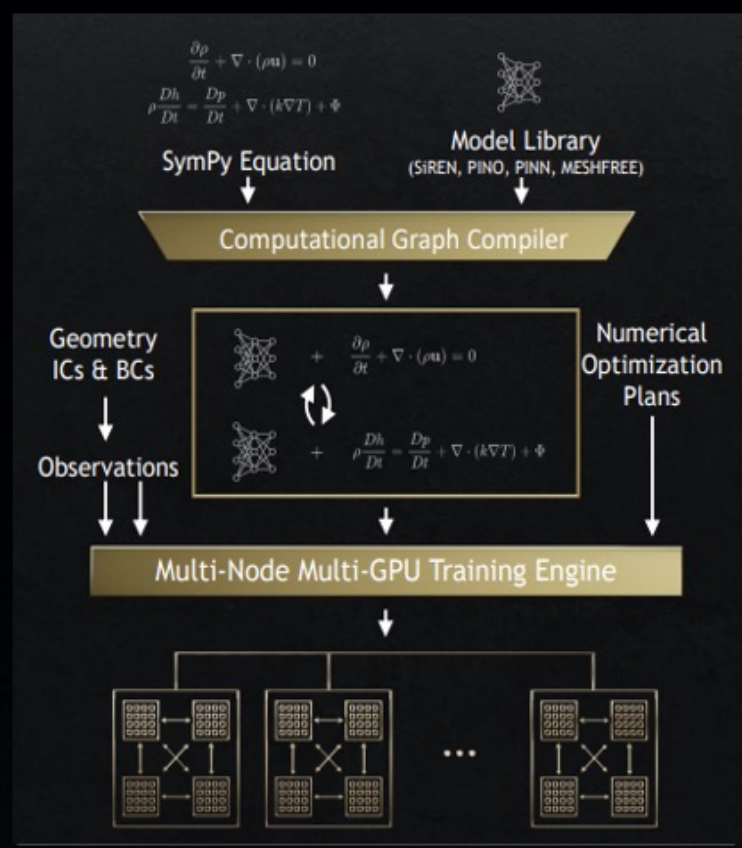
Earth-2 harnesses NVIDIA's full-stack technologies to make Earth digital twins a reality

## Interactive Collaborative Platform



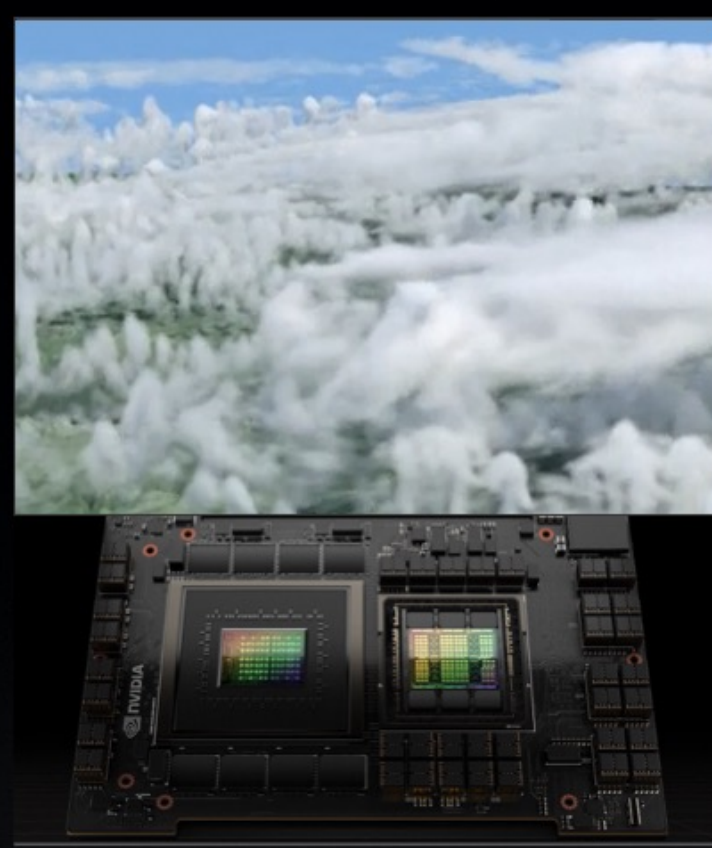
OMNIVERSE

## Data-driven Models



PHYSICS-ML /  
MODULUS

## Storm-resolving Models



GPU-ACCELERATION

## Unified Observations



OMNIVERSE NUCLEUS

## Exascale Compute



OVX SUPERPOD

# Earth-2 is in Collaboration with International Climate Science

NVIDIA's AI, engineering & full-stack expertise complement research capacity in academia & government.



**MAX-PLANCK-INSTITUT FÜR METEOROLOGIE**

**ECMWF**  
EUROPEAN CENTRE FOR MEDIUM RANGE WEATHER FORECASTS

**BSC** **Barcelona Supercomputing Center**  
Centro Nacional de Supercomputación

**CSCS**  
Centro Svizzero di Calcolo Scientifico  
Swiss National Supercomputing Centre

**NASA**

**NOAA**  
NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION  
U.S. DEPARTMENT OF COMMERCE

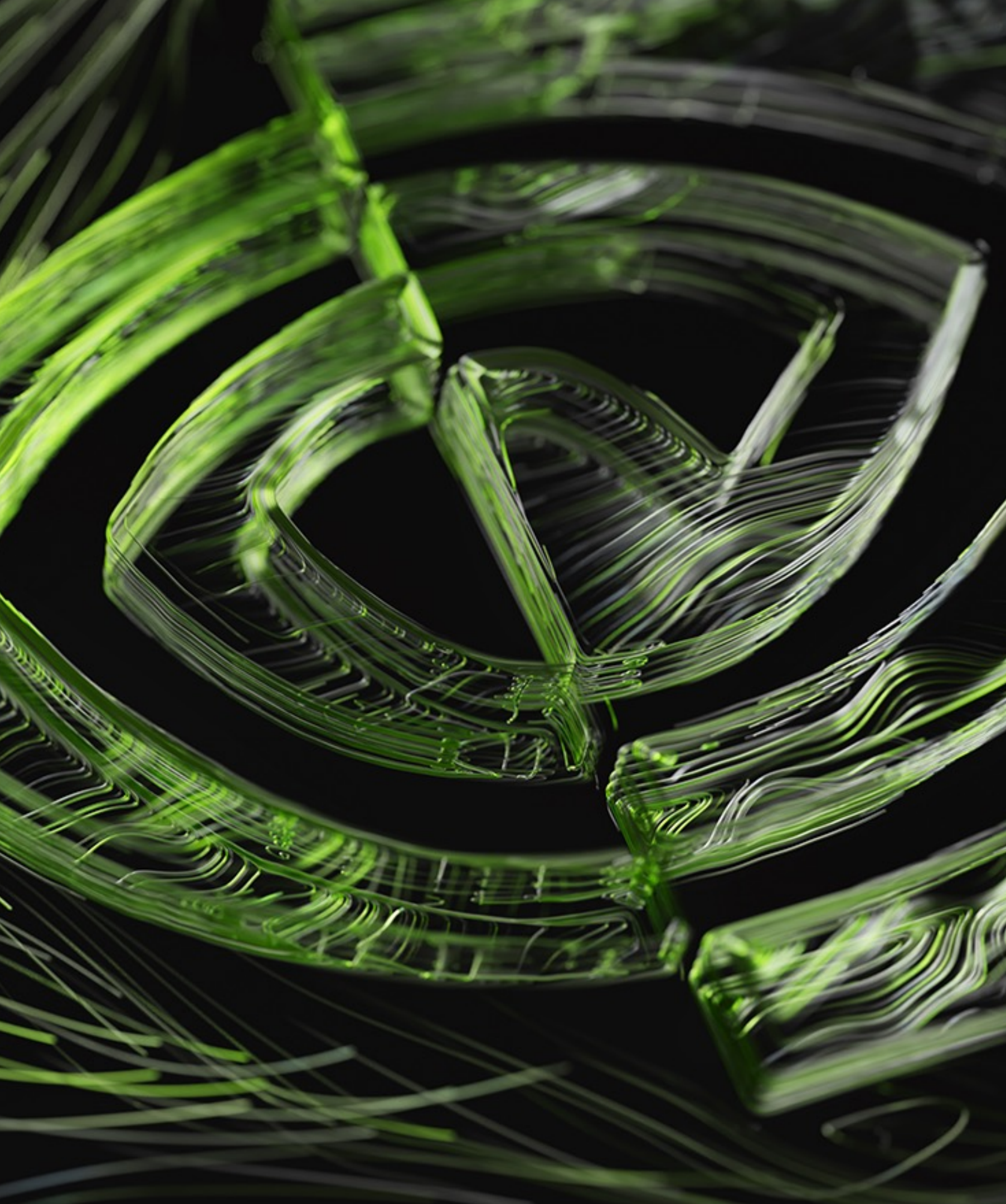
**DEPARTMENT OF ENERGY**  
UNITED STATES OF AMERICA

**USMILE**

**MeteoSwiss**

**Ai2**

**LEAP**



# Agenda

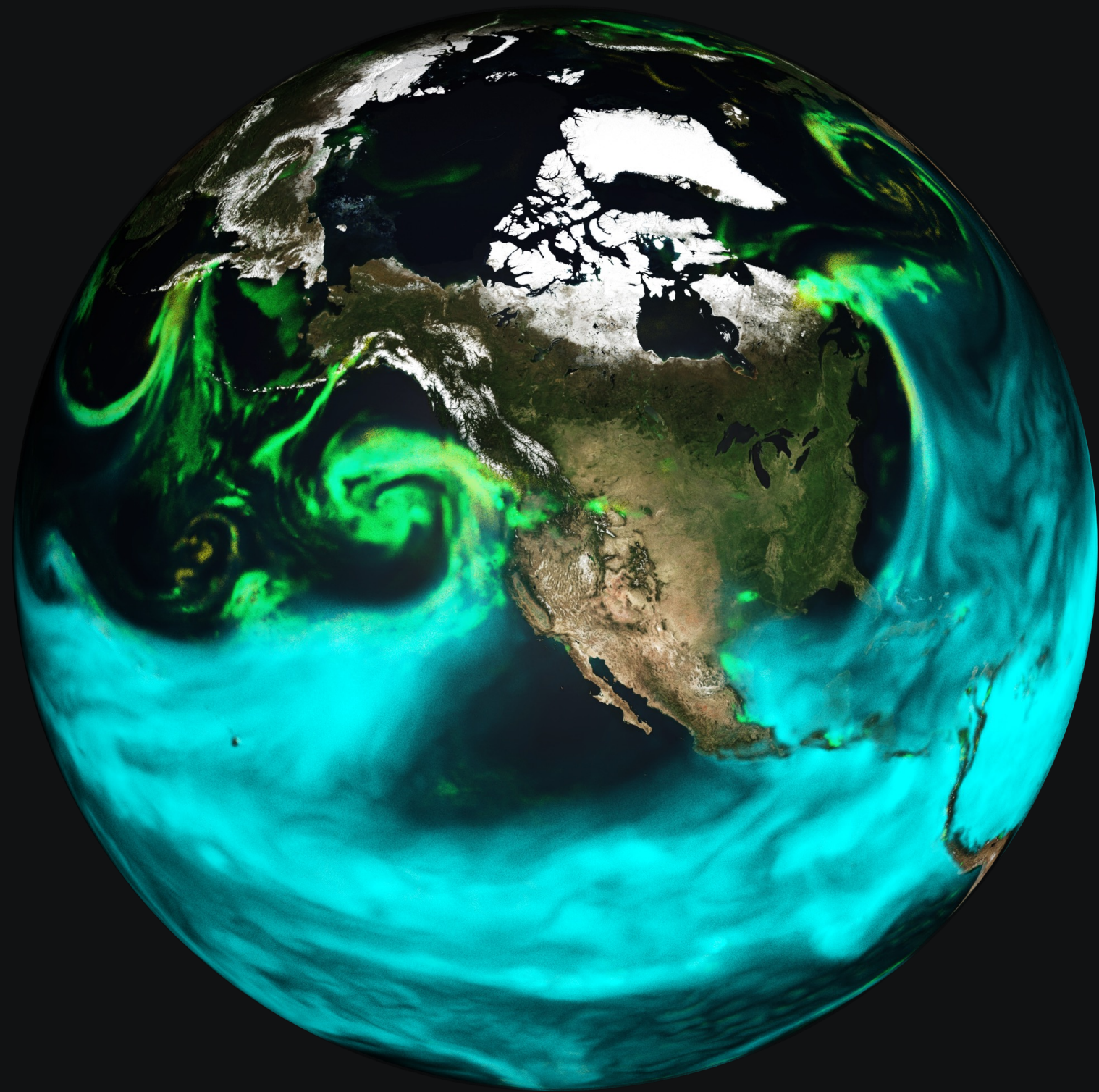
- NVIDIA's Earth-2 initiative: The Big Picture

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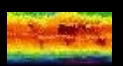
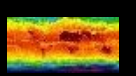
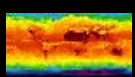
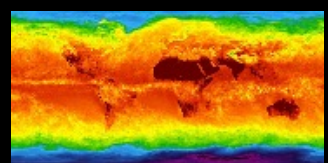
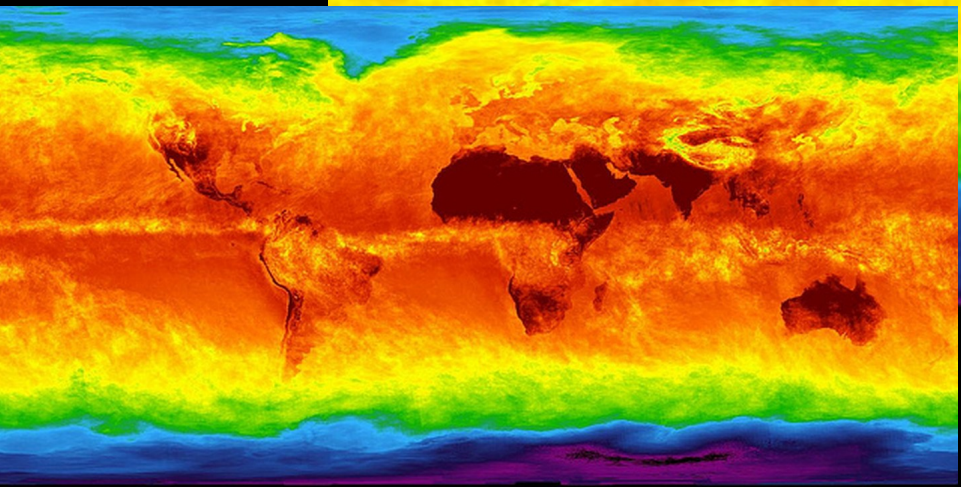
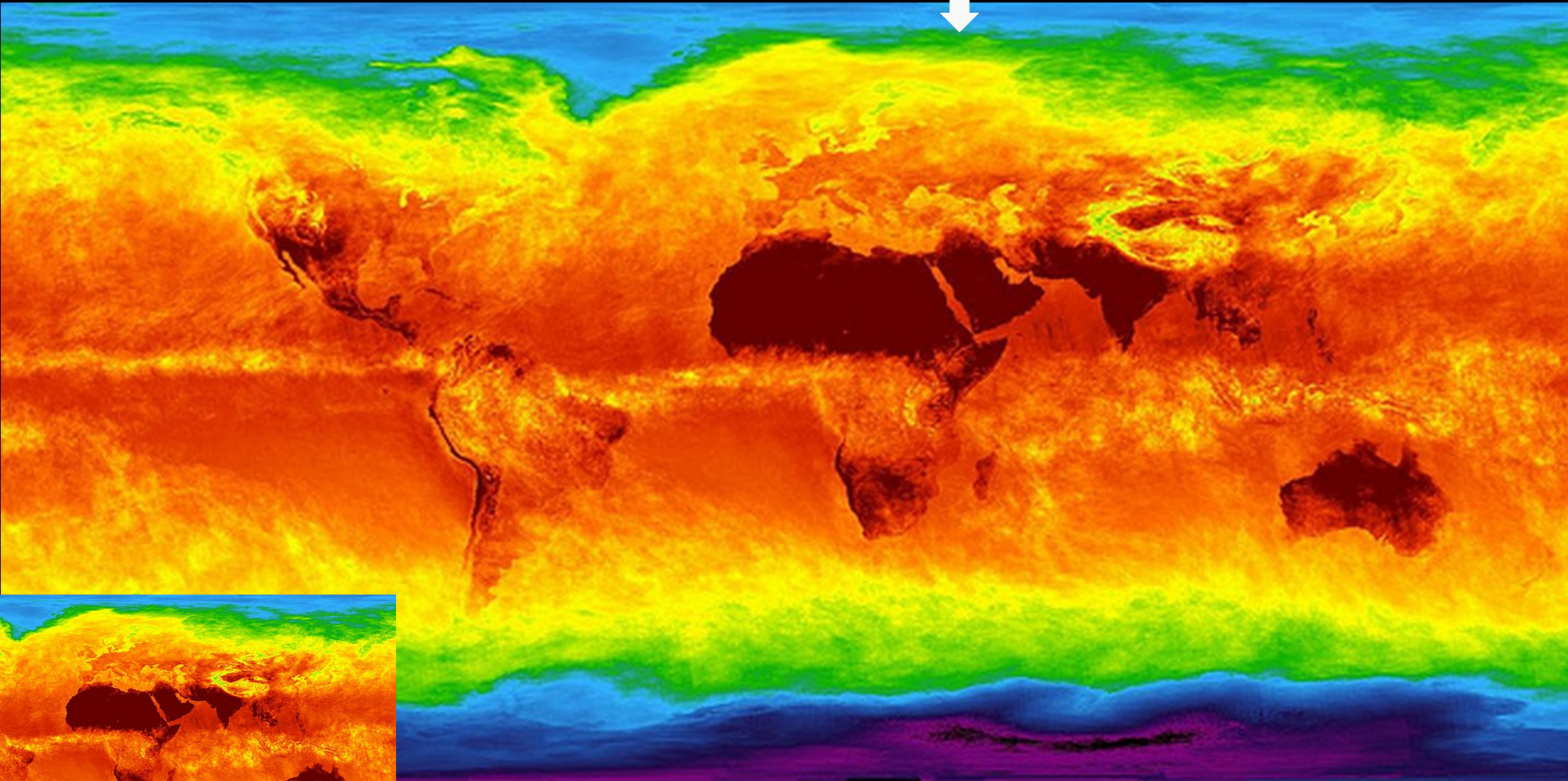
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## FourCastNet pushing the frontier of AI-Driven Digital Twins

|                      |                                       |
|----------------------|---------------------------------------|
| • Scope              | Global                                |
| • Model Type         | Full-Atmosphere AI Surrogate          |
| • Architecture       | Fourier Neural Operator + Transformer |
| • Resolution         | 25km                                  |
| • Training Data      | ERA5 Reanalysis                       |
| • Initial Condition  | ERA5 / GFS / UFS                      |
| • Inference Time     | 0.5 sec (2-week forecast)             |
| • Calibration        | IC + Bayesian model uncertainty       |
| • Speedup vs NWP     | $O(10,000 - 100,000)$                 |
| • Power Savings      | $O(10,000)$                           |
| • Max Stable Rollout | 250+ days                             |
| • Project Type       | Open-source                           |

# FourCastNet: A data-driven weather predictor of **unusually high** resolution



FourCastNet, Pathak et al. (2022), 0.25°, ~1,000,000 Pixels, ViT+AFNO

GNN, Keisler et al. (2022), 1°, 64,000 Pixels, Graph Neural Networks

DLWP, Weyn et al. (2020). 2°, 16K pixels, Deep CNN on Cubesphere/(2021) ResNet

Weyn et al. (2019), 2.5° N.H only, 72x36, 2.6k pixels, ConvLSTM

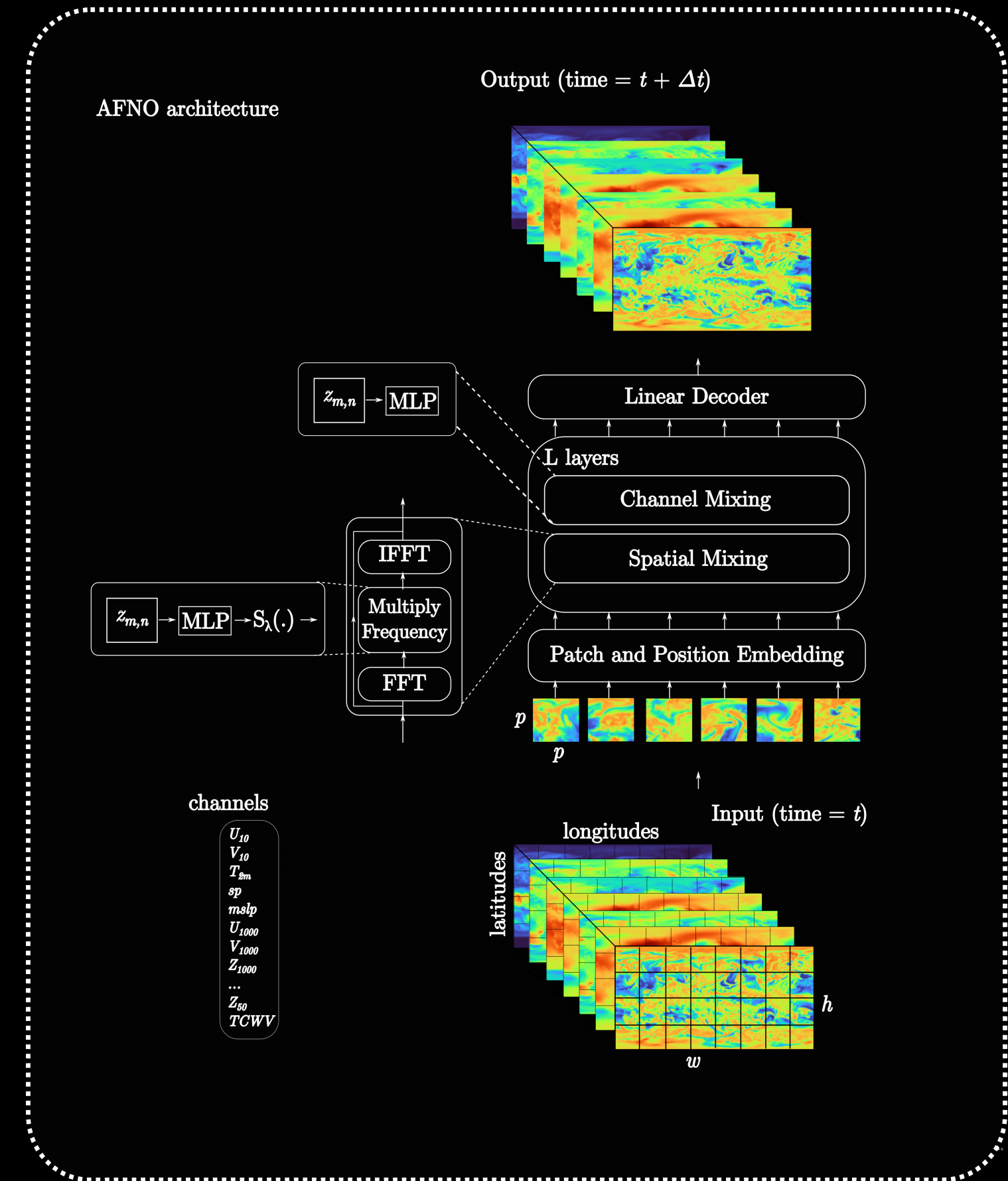
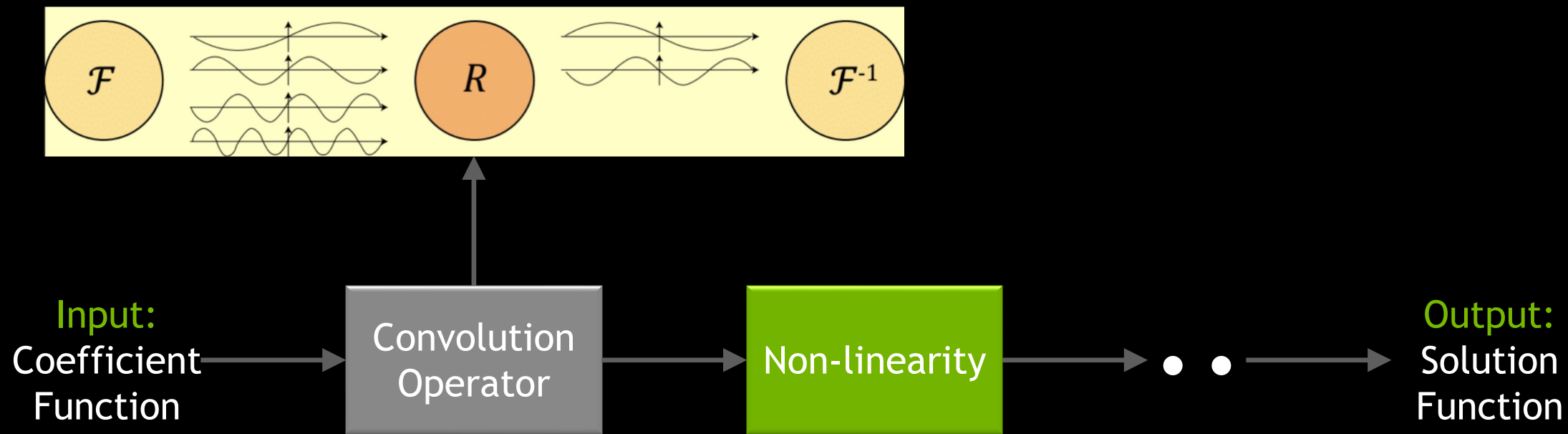
WeatherBench, Rasp et al. (2020). 5.625°, 64x32, 2K pixels, CNN

Deuben & Bauer (2018), 6° , 60x30, 1.8K pixels, MLP



# FourCastNet uses a novel transformer architecture

With Fourier Neural Operator Blocks - in search of grid-free, high-resolution, machine-learnt simulations.



# FourCastNet (FCN) is trained on 0.25-degree ERA5 data

With 26 channels (2D fields) of surface and atmospheric variables.

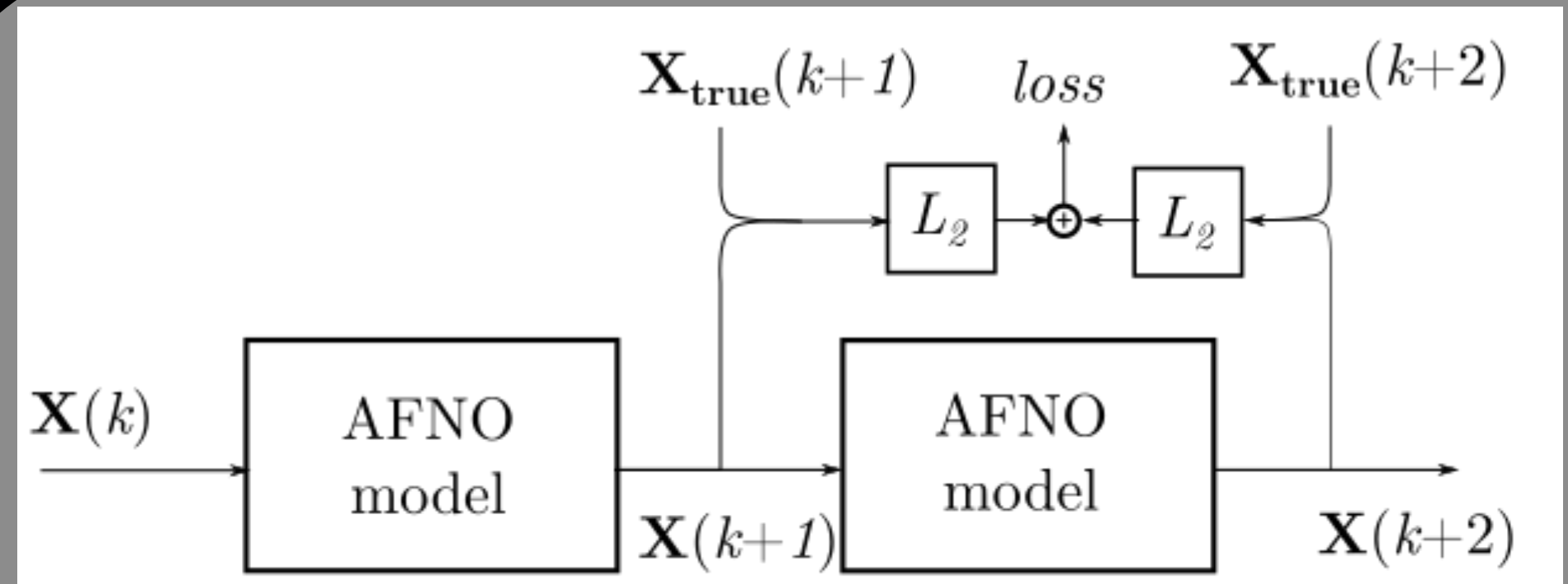
- Surface: U, V, T, MSLP
- 5 vertical levels: U, V, T, Z, RH
- Integrated column water vapor

Extending to include radiation, surface and TOA fluxes, vapor transport, clouds

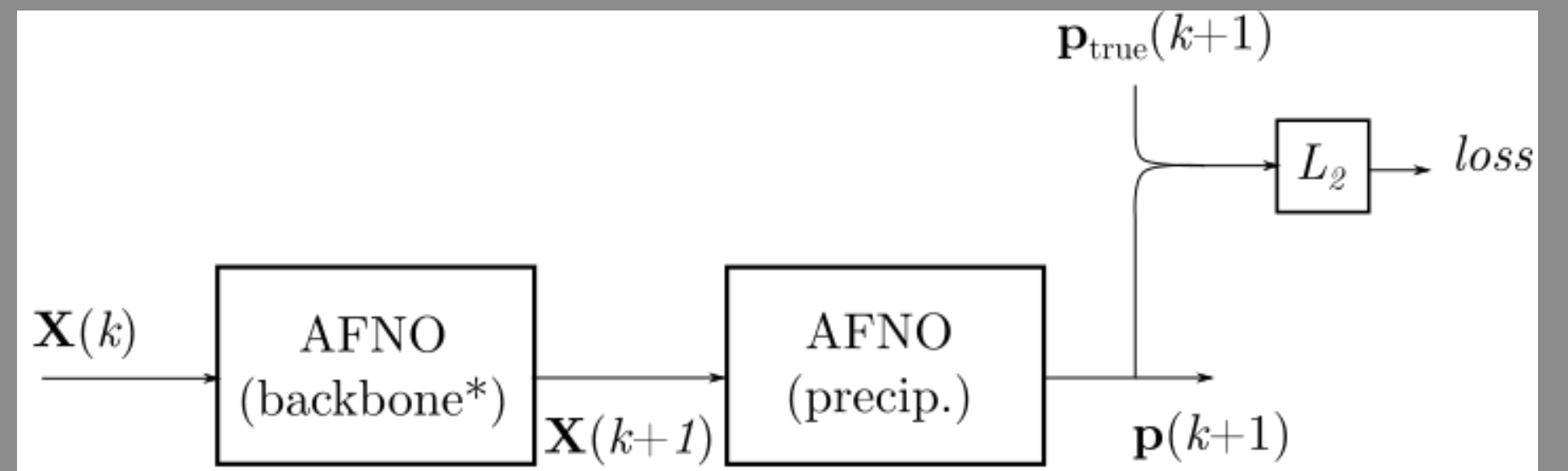
Training set: 1979 to 2015

Validation set: 2016, 2017

Held out: 2018 onwards



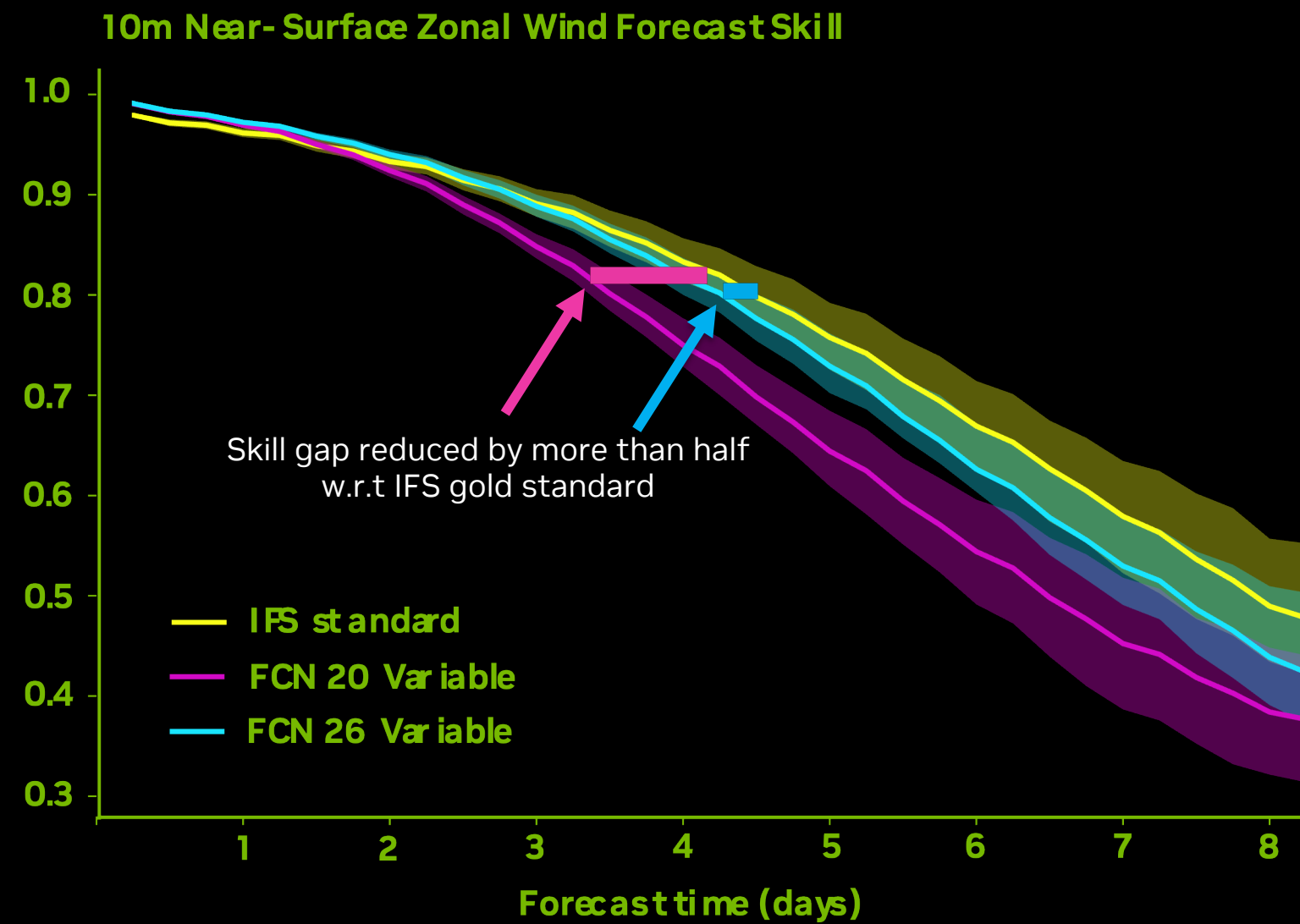
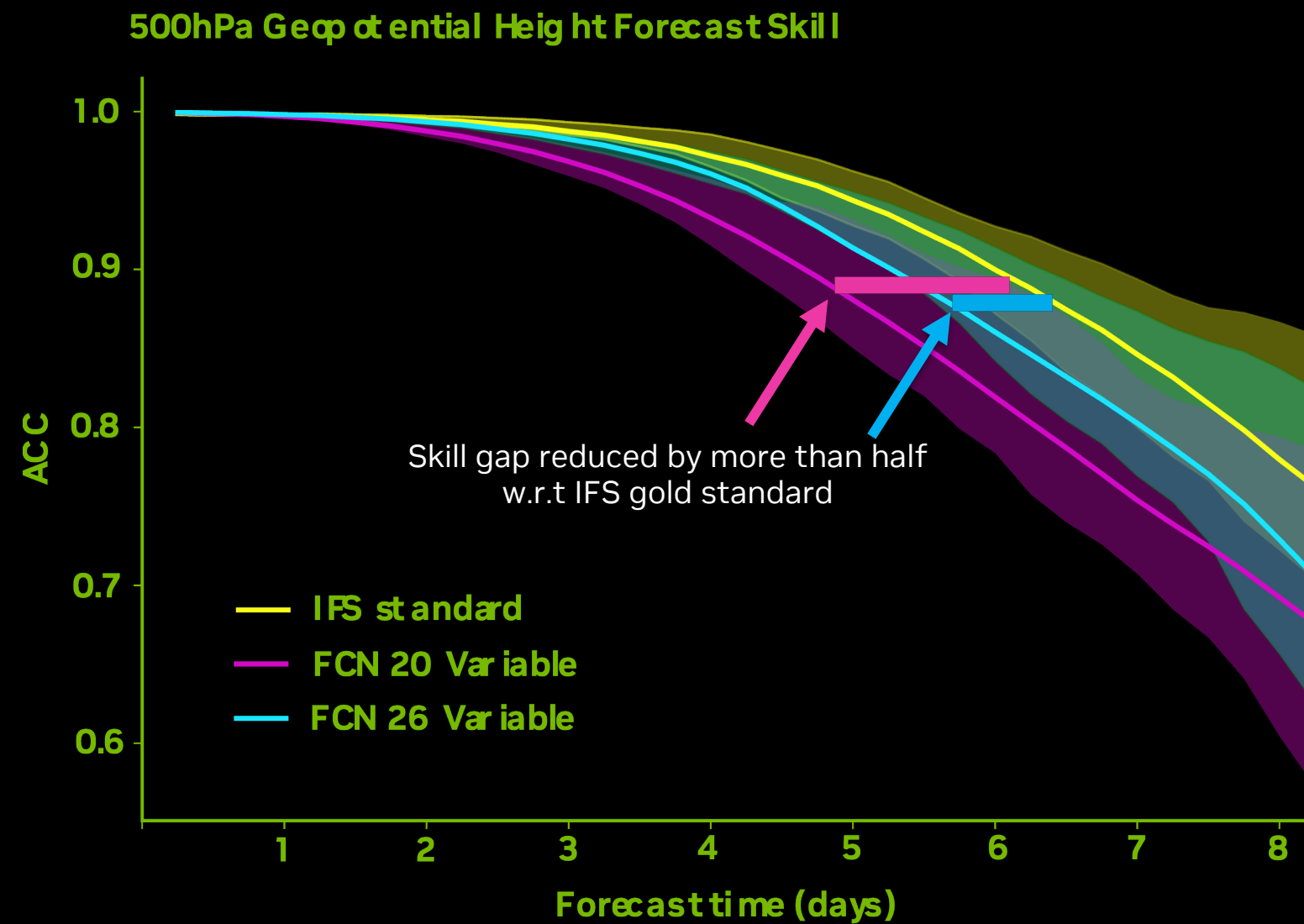
Two step fine-tuning



Diagnosed precip

# FCN medium-range weather forecast skill improving with training ambition.

Could it one day outperform deterministic models? We don't yet know the limit.



## Acronym Alert:

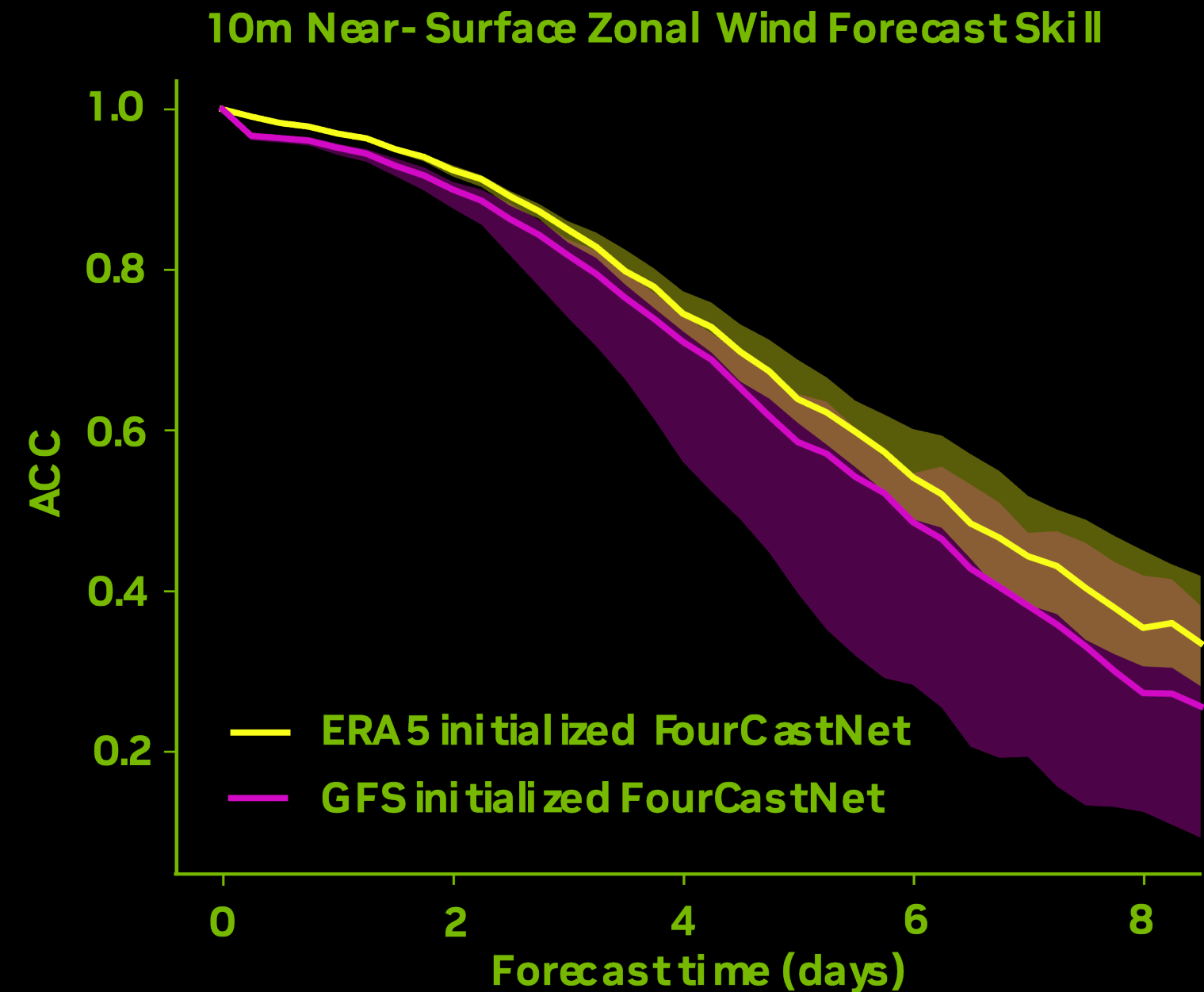
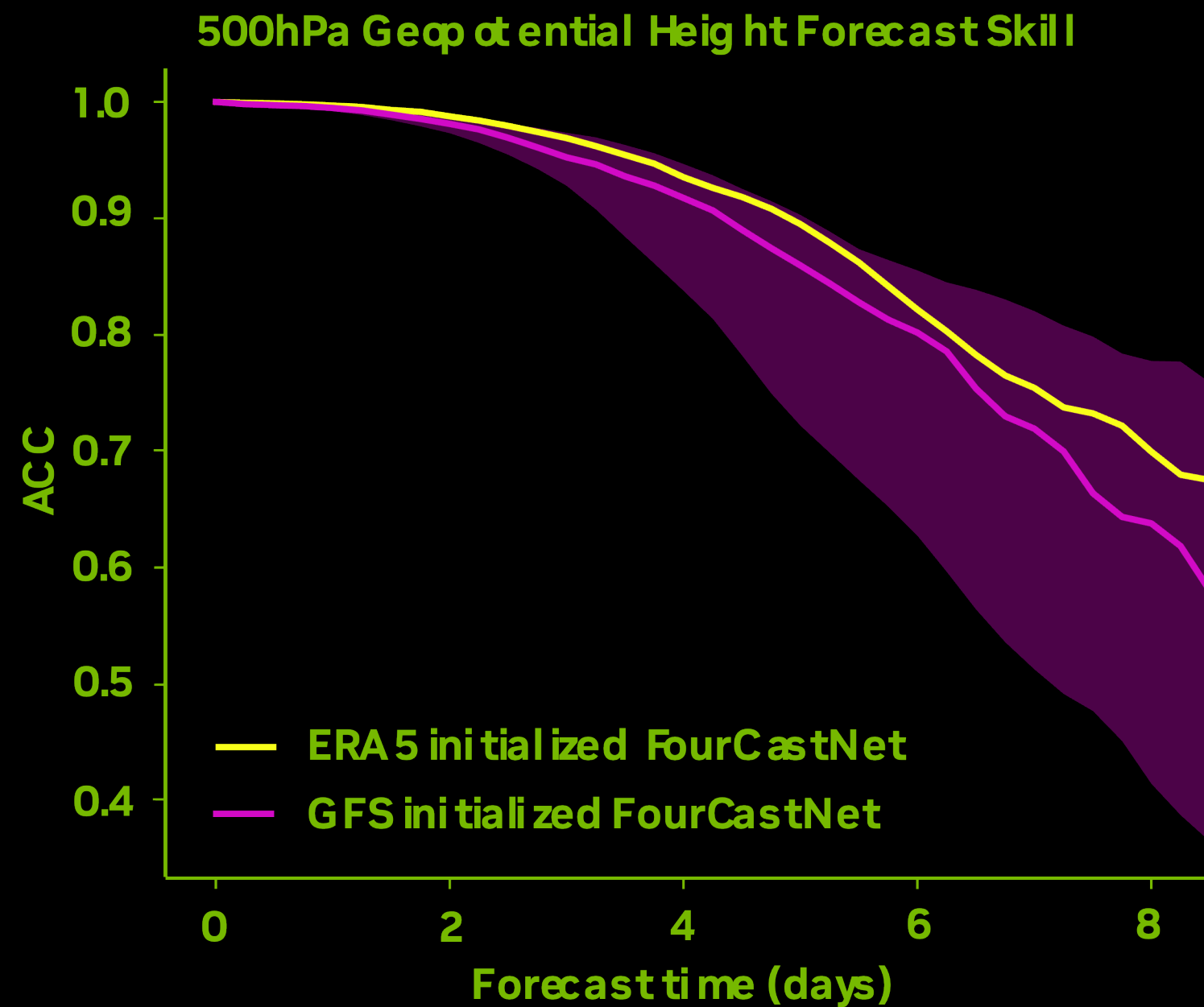
**ACC:** Anomaly Correlation Coefficient (metric of weather skill)

**IFS:** The Integrated Forecast System, a gold standard weather model

**FCN:** FourCastNet, our digital twin of weather.

# Can FCN be initialized with real-time conditions?

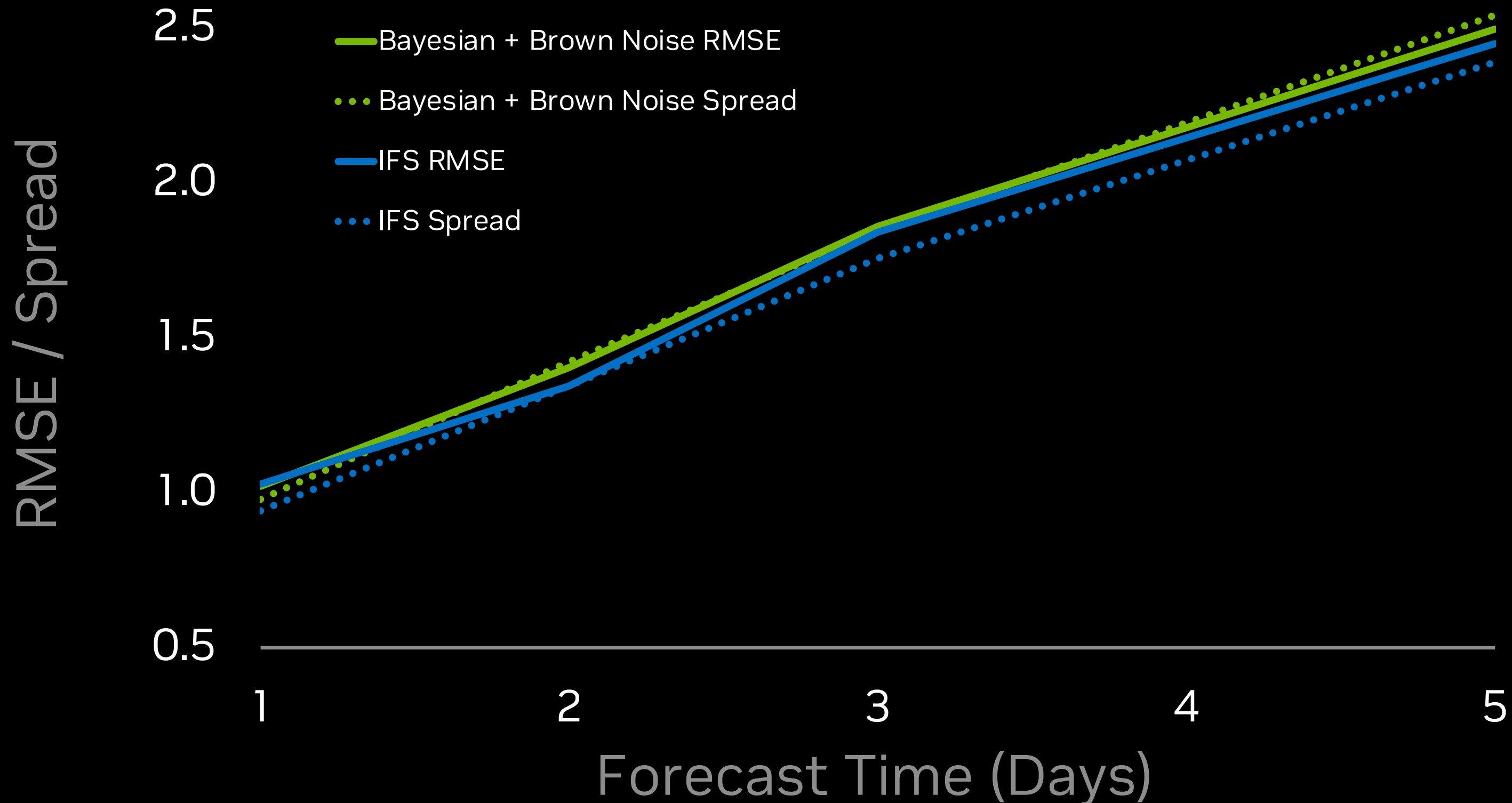
Yes. Zero-shot skill transfer using initial conditions from a separate US dataset that FCN was not directly trained on.



# Probabilities: Spread matters as much as Skill

FCN's ensembles calibrated using initial condition uncertainty and model uncertainty (Bayesian SWA-G).

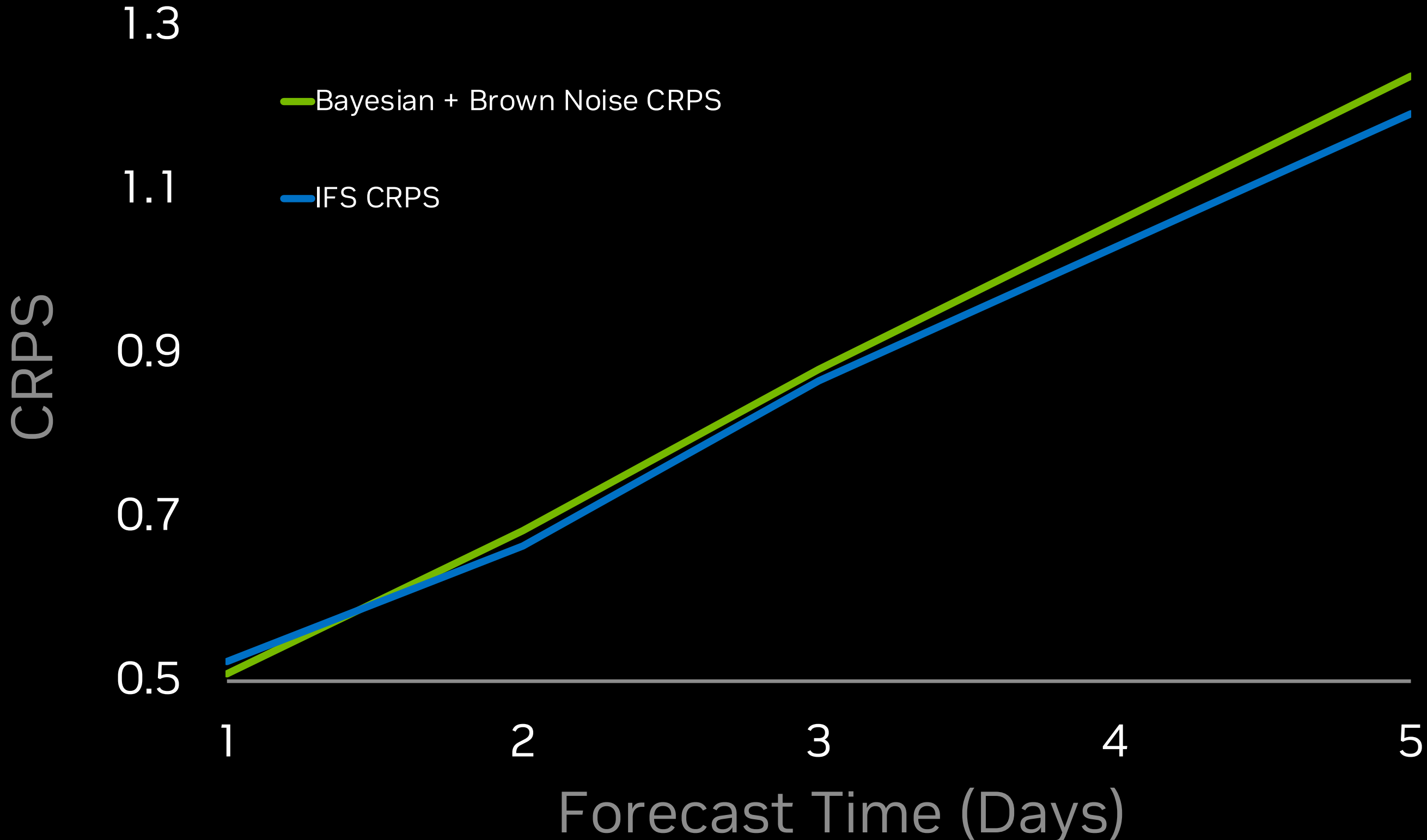
U-component of wind-speed at 10m



# Continuous Ranked Probability Score competitive with IFS standard

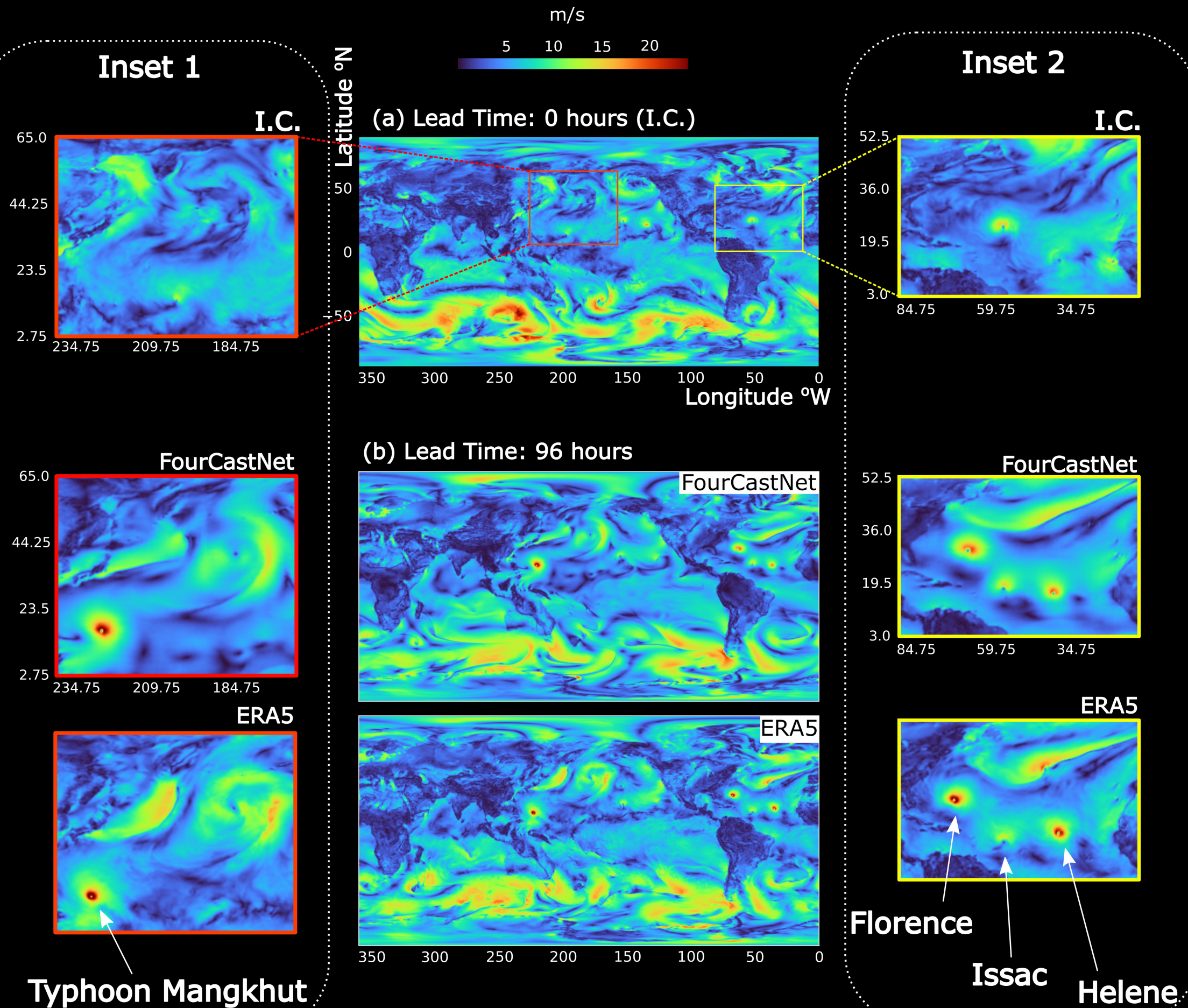
FCN's ensembles calibrated using initial condition uncertainty and model uncertainty (Bayesian SWA-G).

U-component of wind-speed at 10m



# FCN has impressive skill on forecasting extremes.

Including tropical cyclones, extra-tropical cyclones, and atmospheric rivers.

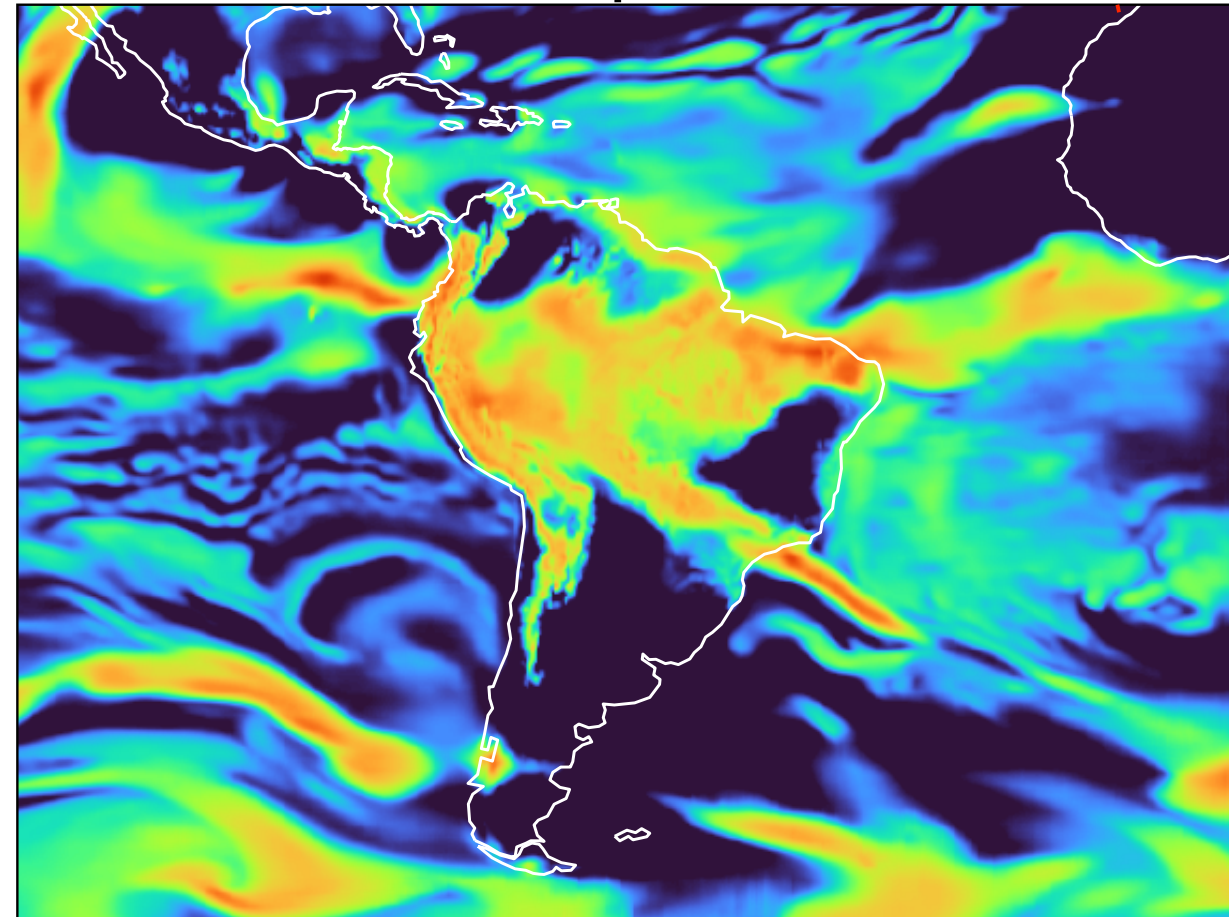


# Can generative approaches improve regional extremes?

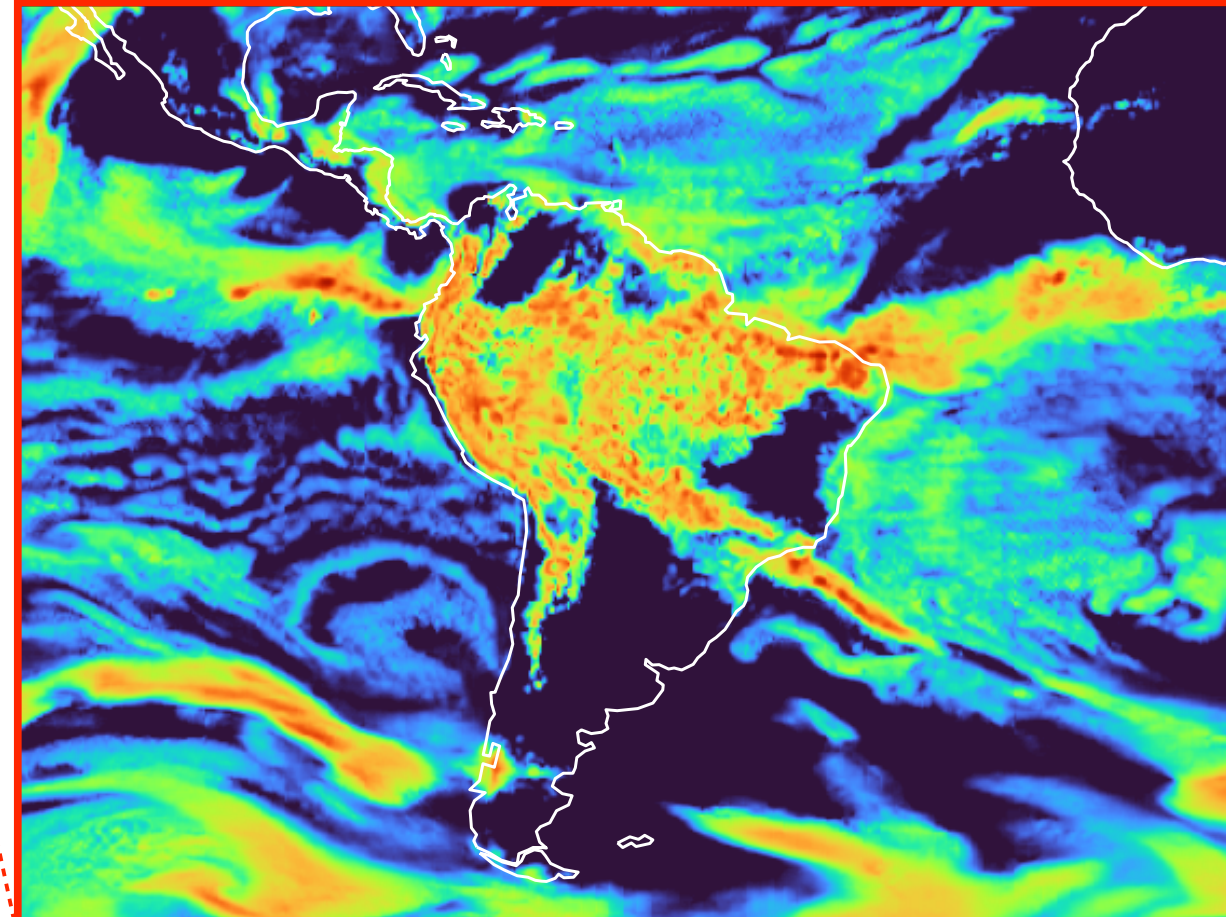
Yes. Adversarial loss improves fine scale detail skillfully.

Forecast lead time of 18 hours

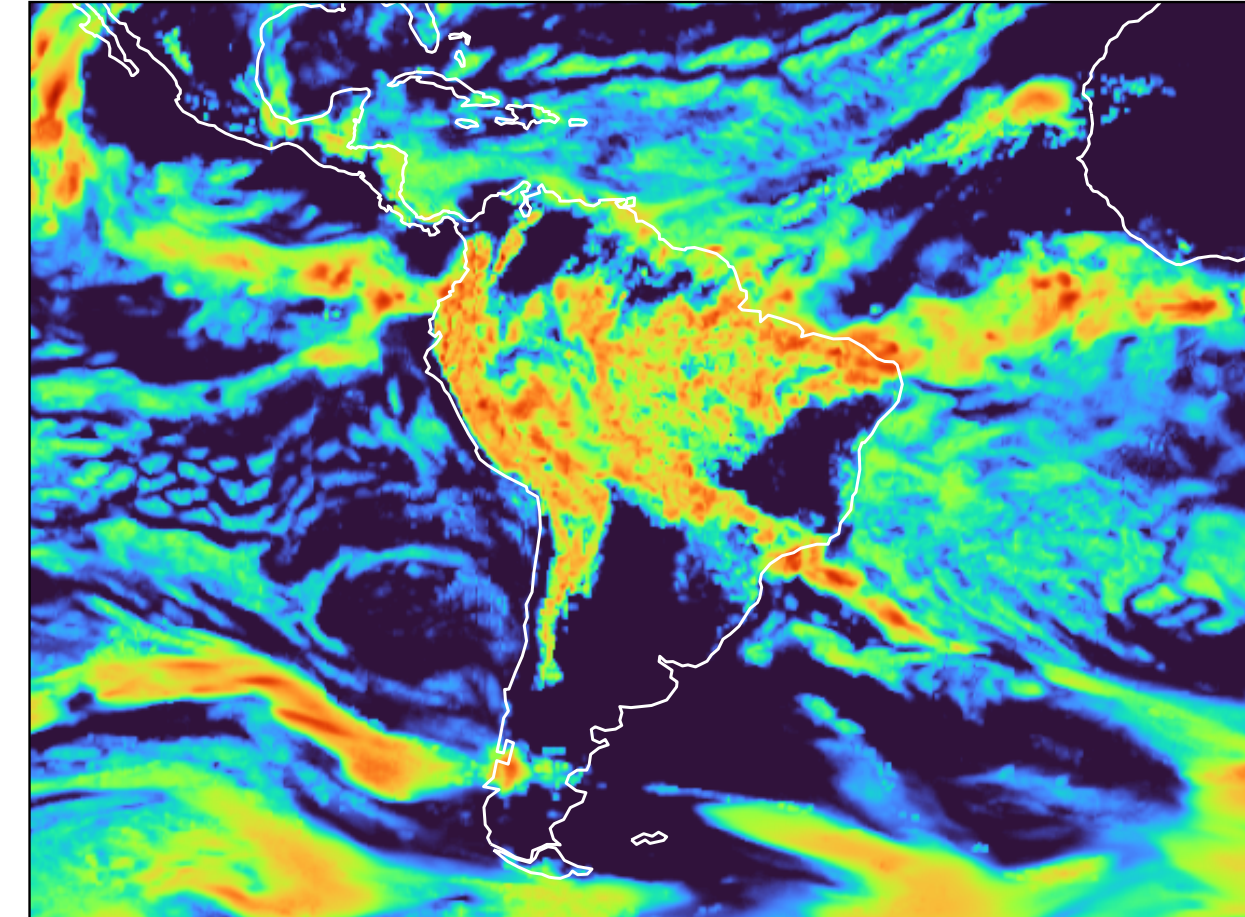
L1 loss (previous)



L1 + adversarial



ERA5 ground truth

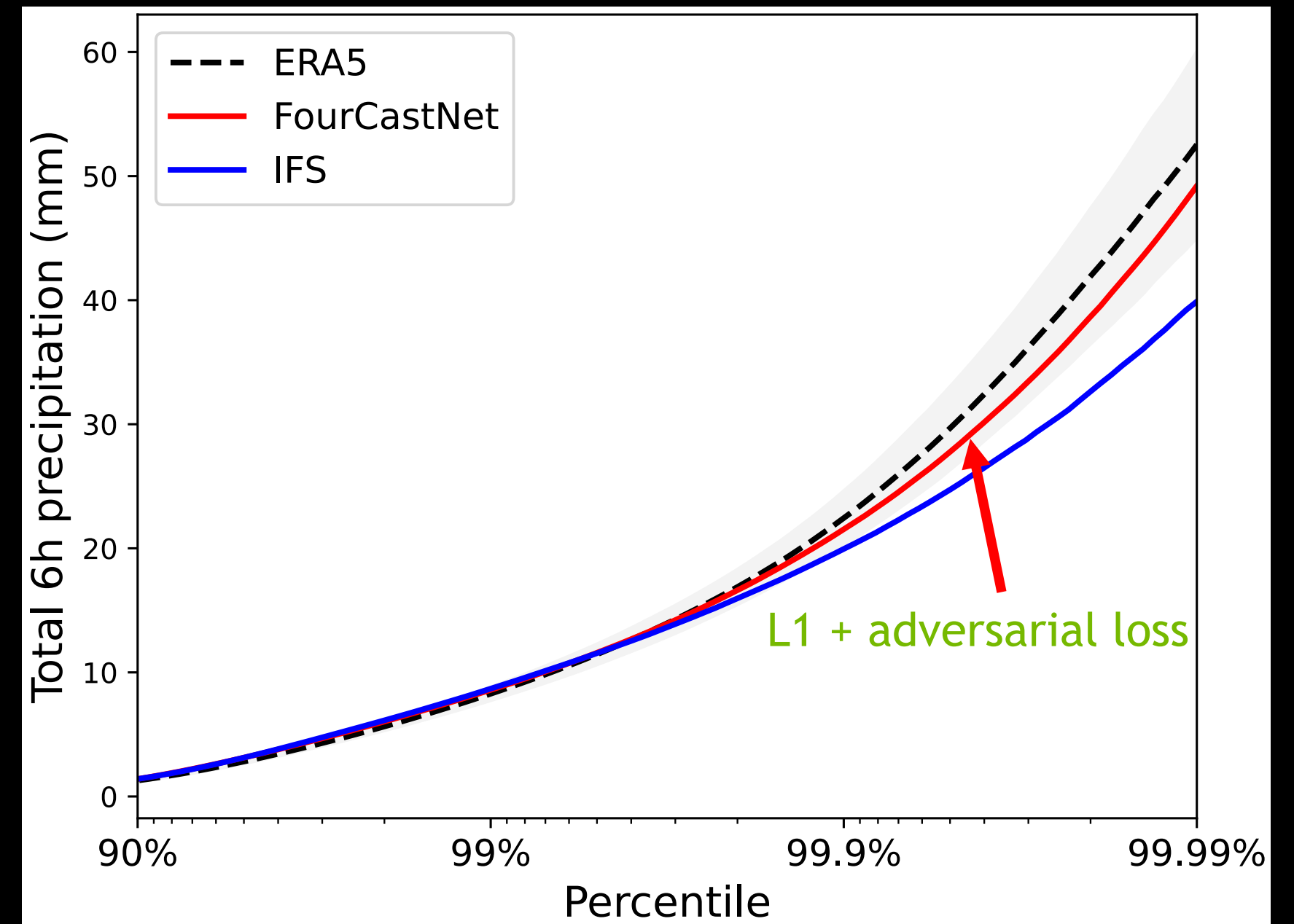
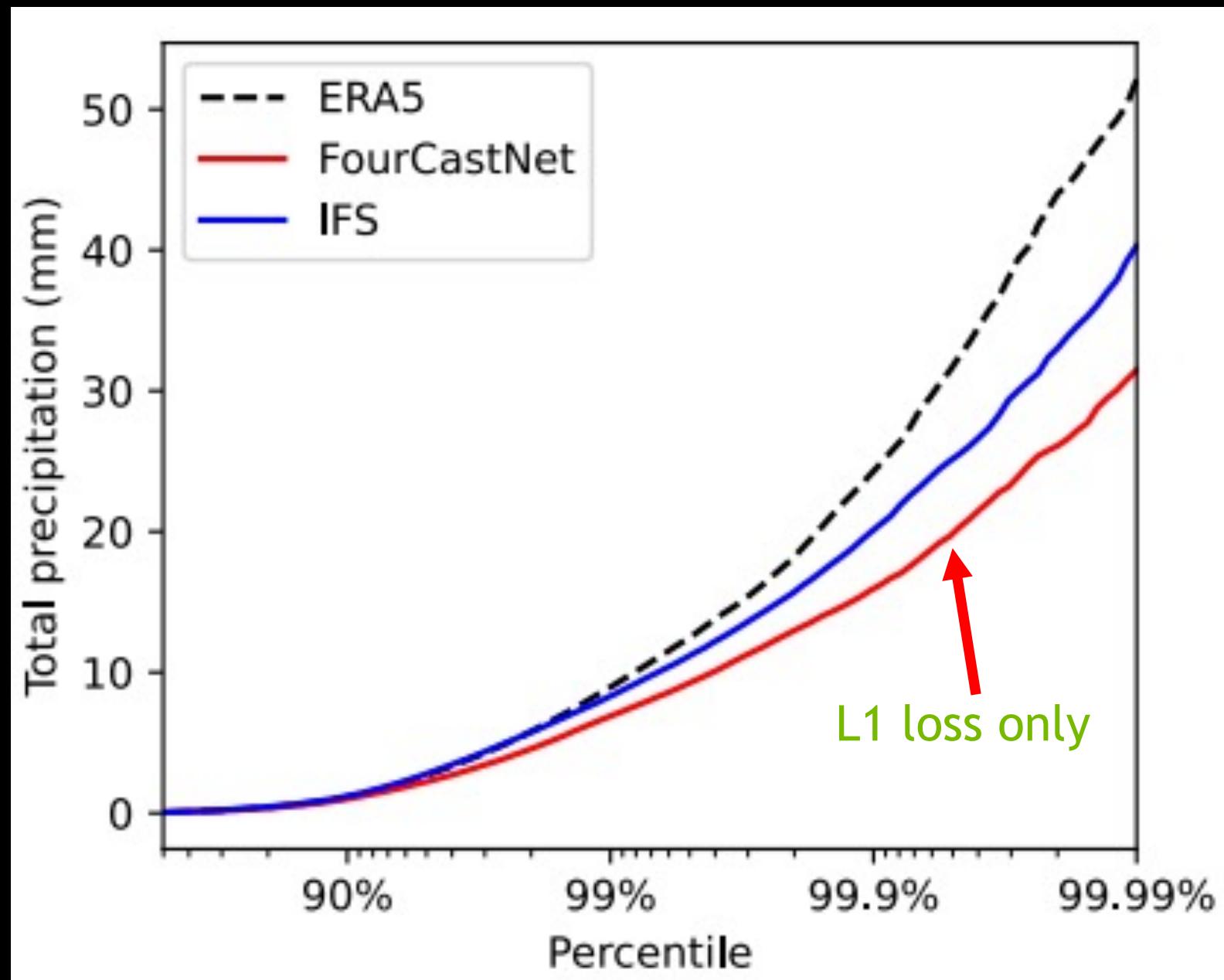




# Progress in capturing extreme precipitation statistics

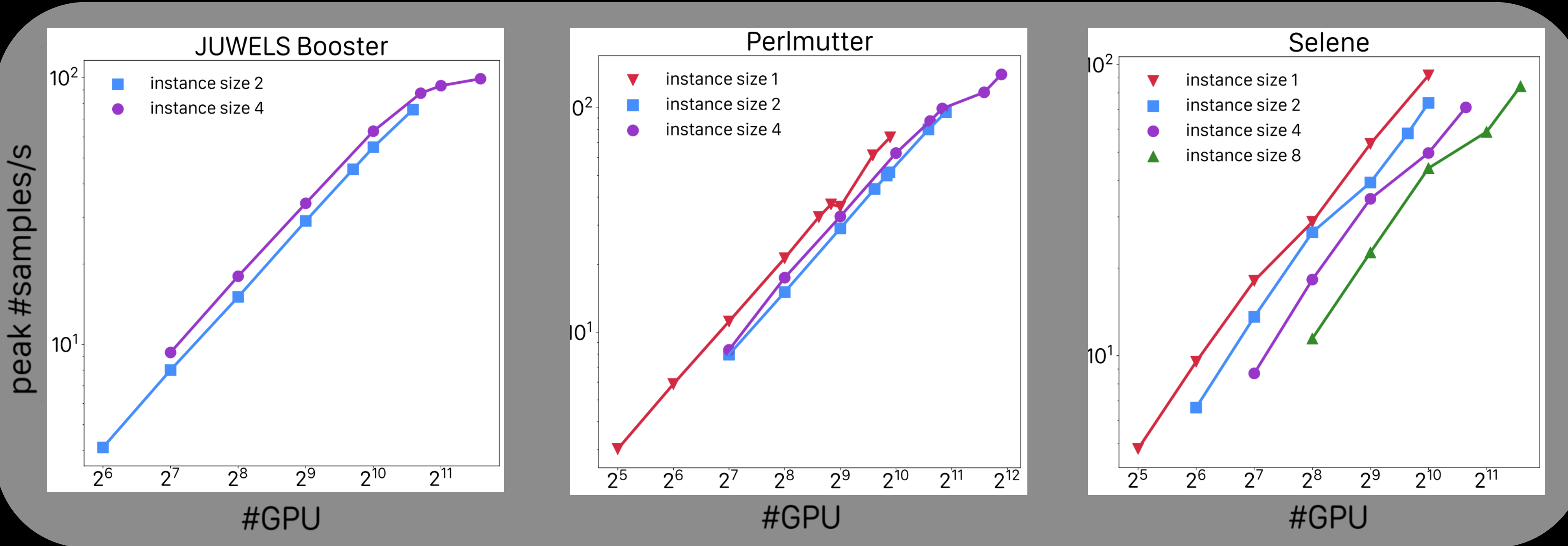
Adding generative adversarial loss improves predictions of rarest, most intense rainfall events.

Forecast lead time of 18 hours



# FCN trained on ambitious amounts of data scales efficiently up to ~ 4000 GPUs on three supercomputing systems

Thanks to full-stack AI + HPC expertise we train on a growing amount of the world's petabytes of past weather data.



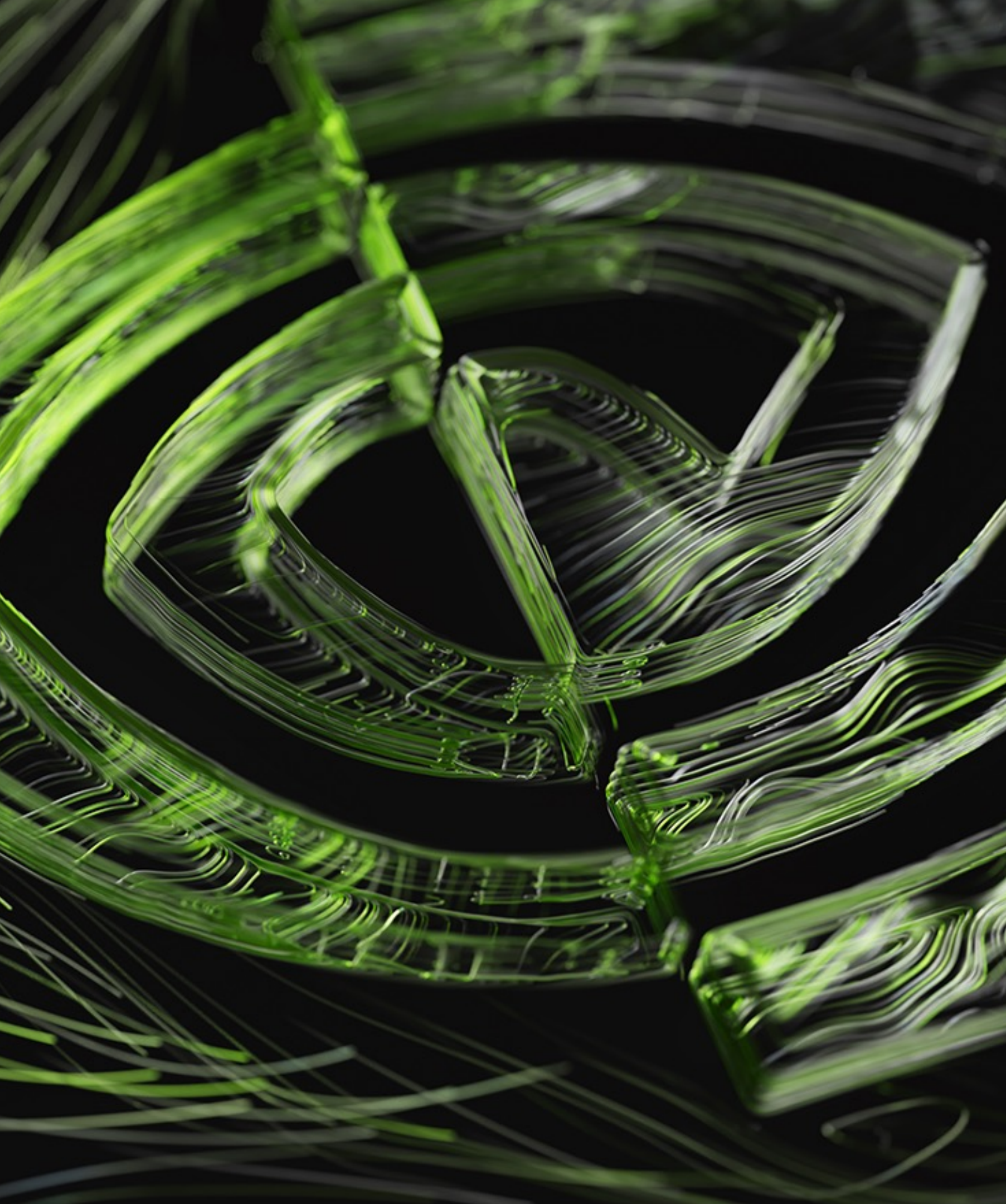
Peak performance is 140.8 petaFLOPS in mixed precision (averaged over a full epoch).  
Time to solution decreased from 24+ hours to 67 minutes with model and data parallelism

# FCN's 50,000x speedup w.r.t NWP enables massive ensembles in seconds

With over 10,000x smaller energy footprint

## Computational and energy costs of 100-member ensemble forecast

|                          | IFS (18km) | FCN (25km) | IFS / FCN |
|--------------------------|------------|------------|-----------|
| • Nodes Required         | 3,060      | 1          | 1530      |
| • Latency (node-seconds) | 984,000    | 7          | 44727     |
| • Energy consumed (kJ)   | 271,000    | 7          | 12318     |



# Agenda

- NVIDIA's Earth-2 initiative: The Big Picture

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- FourCastNet & latest accomplishments in Weather.

---
- **Beyond Weather, towards Climate.**

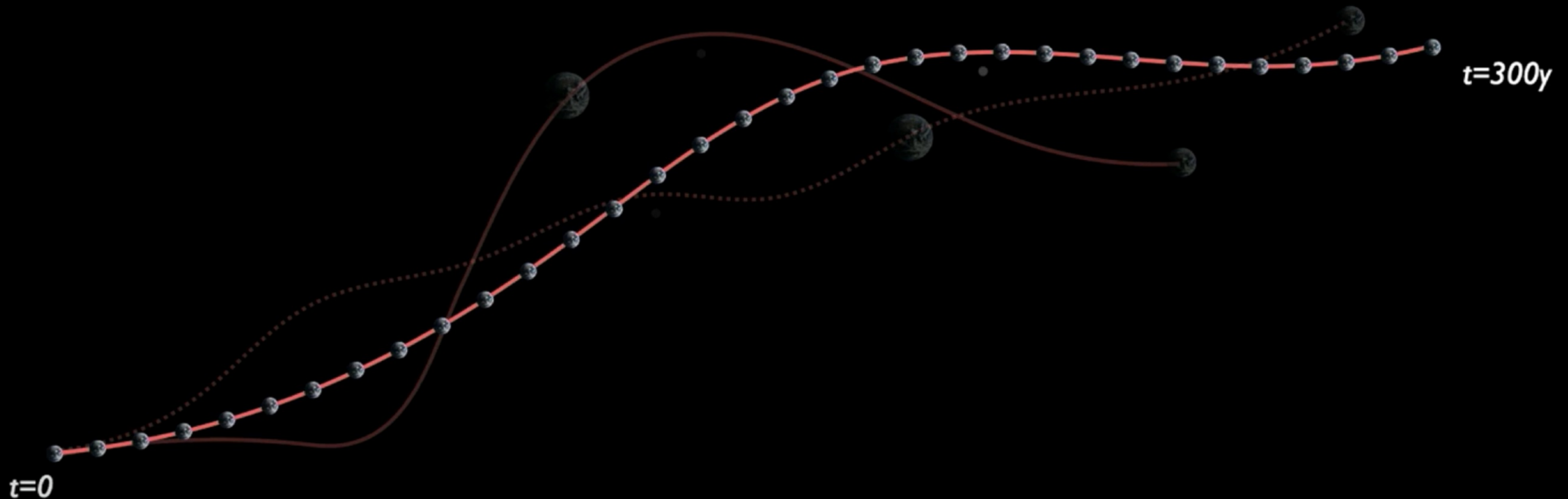
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- Digital Twinning platforms

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# Given Future Data, FourCastNet's Speed Allows Fast Tethering

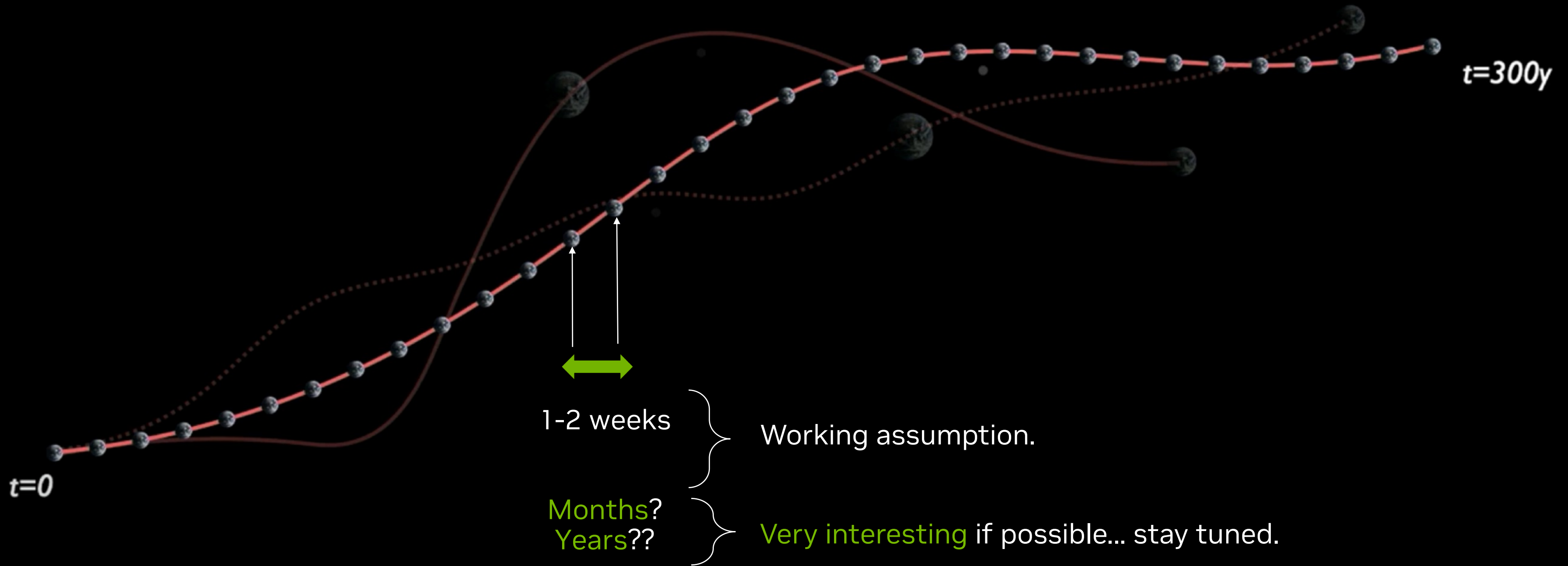
AI nimbly generates details between "checkpoints" saved only infrequently from physics-based climate simulations

-- Bjorn Stevens, GTC 2021

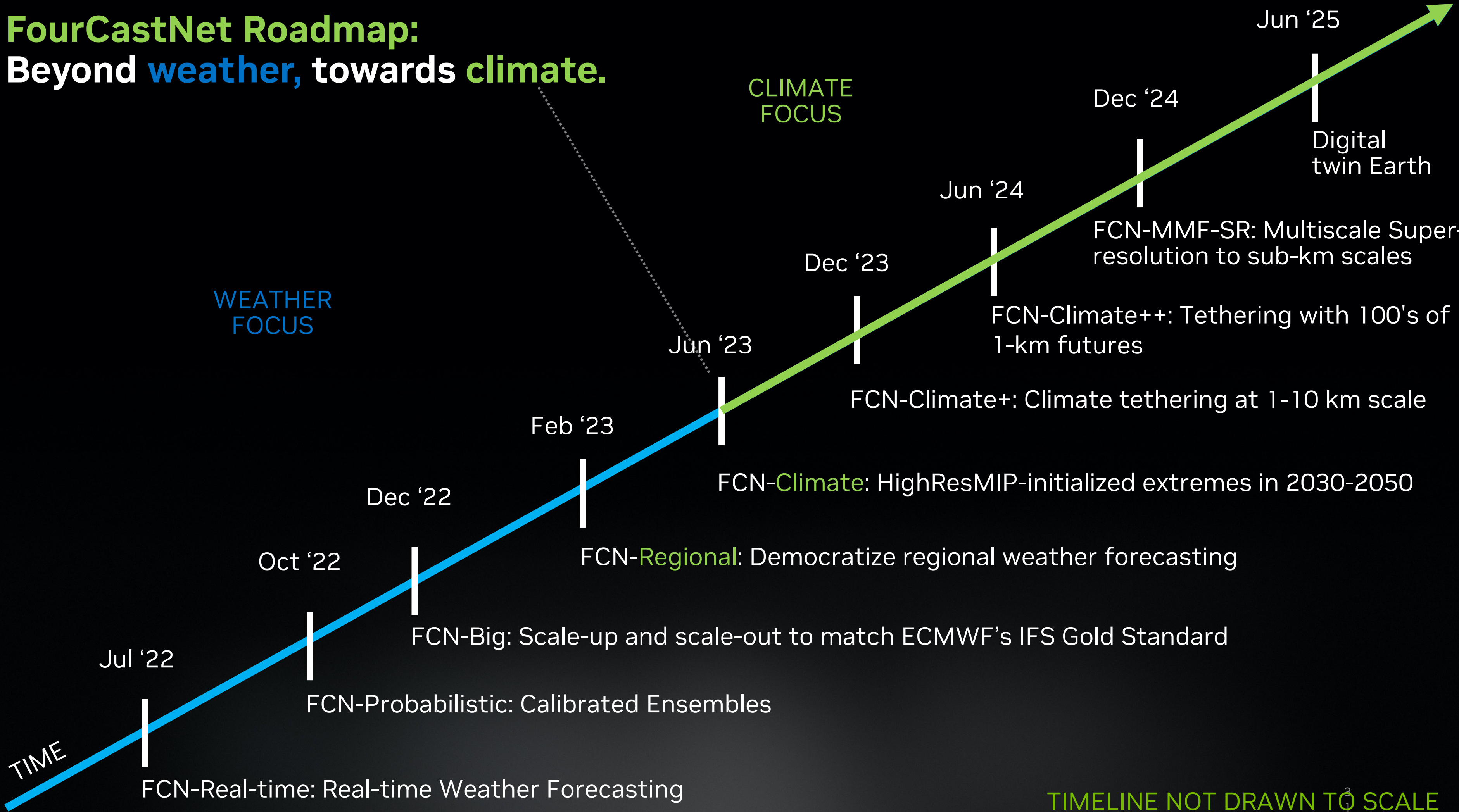


# Open Question

For how long can full-AI models like FourCastNet be trusted to "tether" between climate checkpoints?



# FourCastNet Roadmap: Beyond weather, towards climate.



TIMELINE NOT DRAWN TO SCALE

# To Begin, we can Tether to Existing Climate Predictions.

Using the world's current data library of 25-km resolution HighResMIP climate predictions.

## Sampling scenarios

Mitigation

Population

Pollution

Land-use

Physical projections of  
forced change &  
natural variability

Data-driven  
Tethering

Digital  
Twin

Renewable  
Energy  
Forecasting

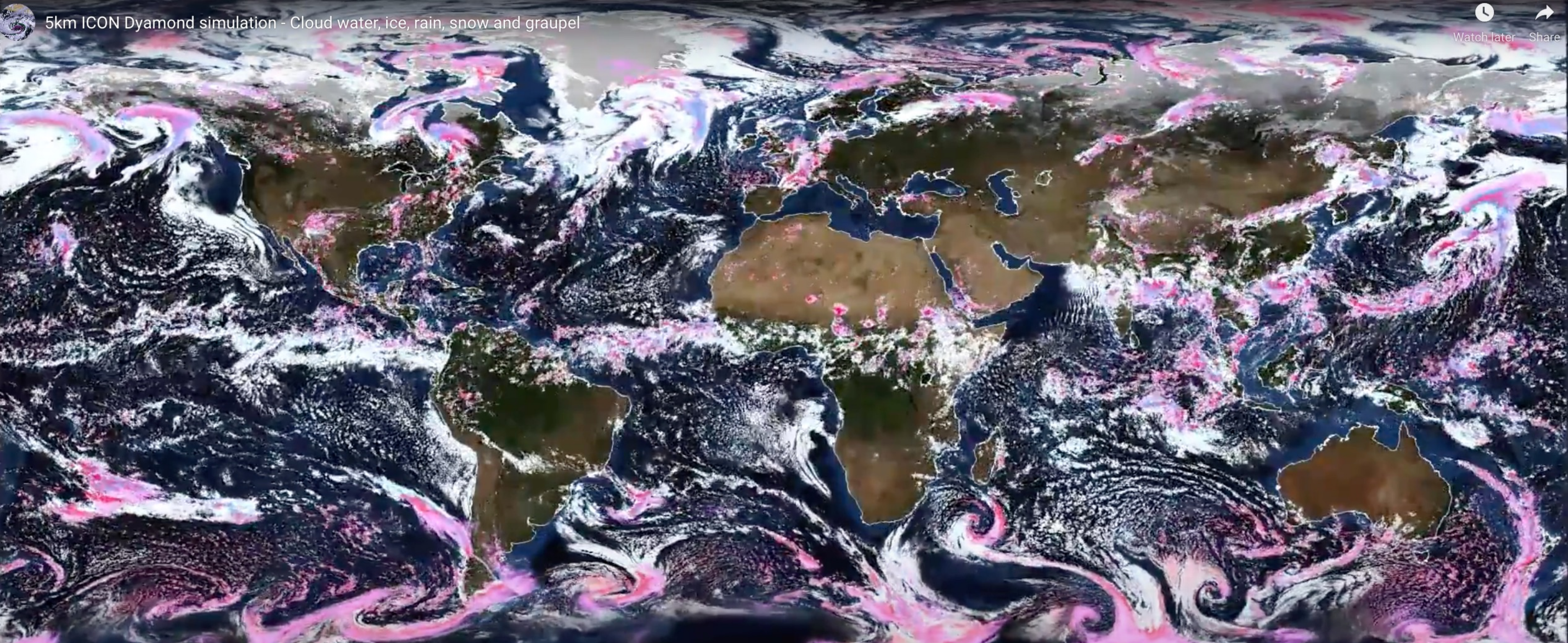
Extreme  
Weather  
Prediction

Disaster  
Mitigation



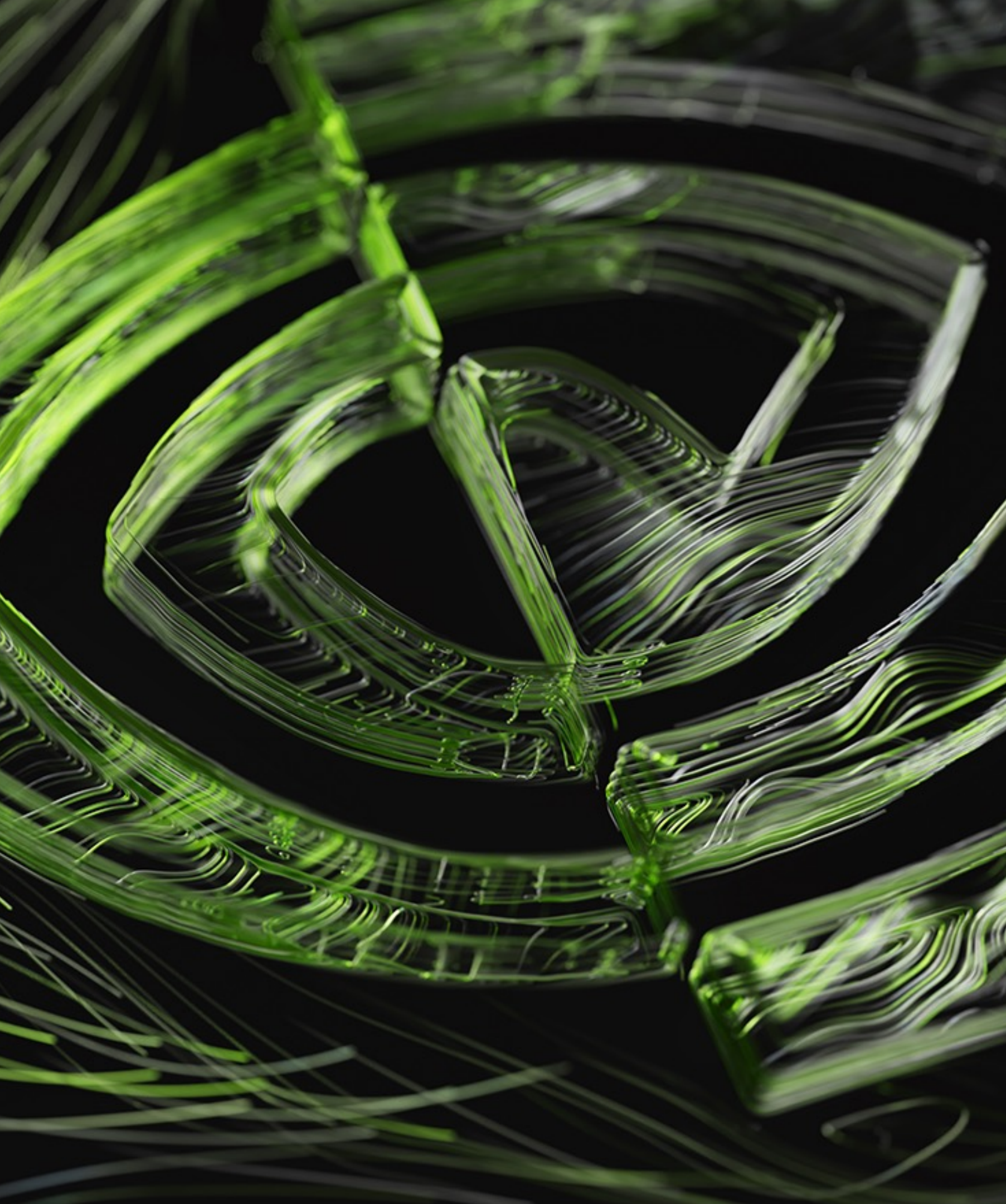
# But Eventually we want to Tether to km-scale Predictions.

Because credible cloud feedbacks and storm dynamics from km-scale simulators matter to predicting regional risk.



5km ICON Dyamond simulation - Cloud water, ice, rain, snow and graupel

Watch later Share



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# Nimbus will enable end-to-end MLOps

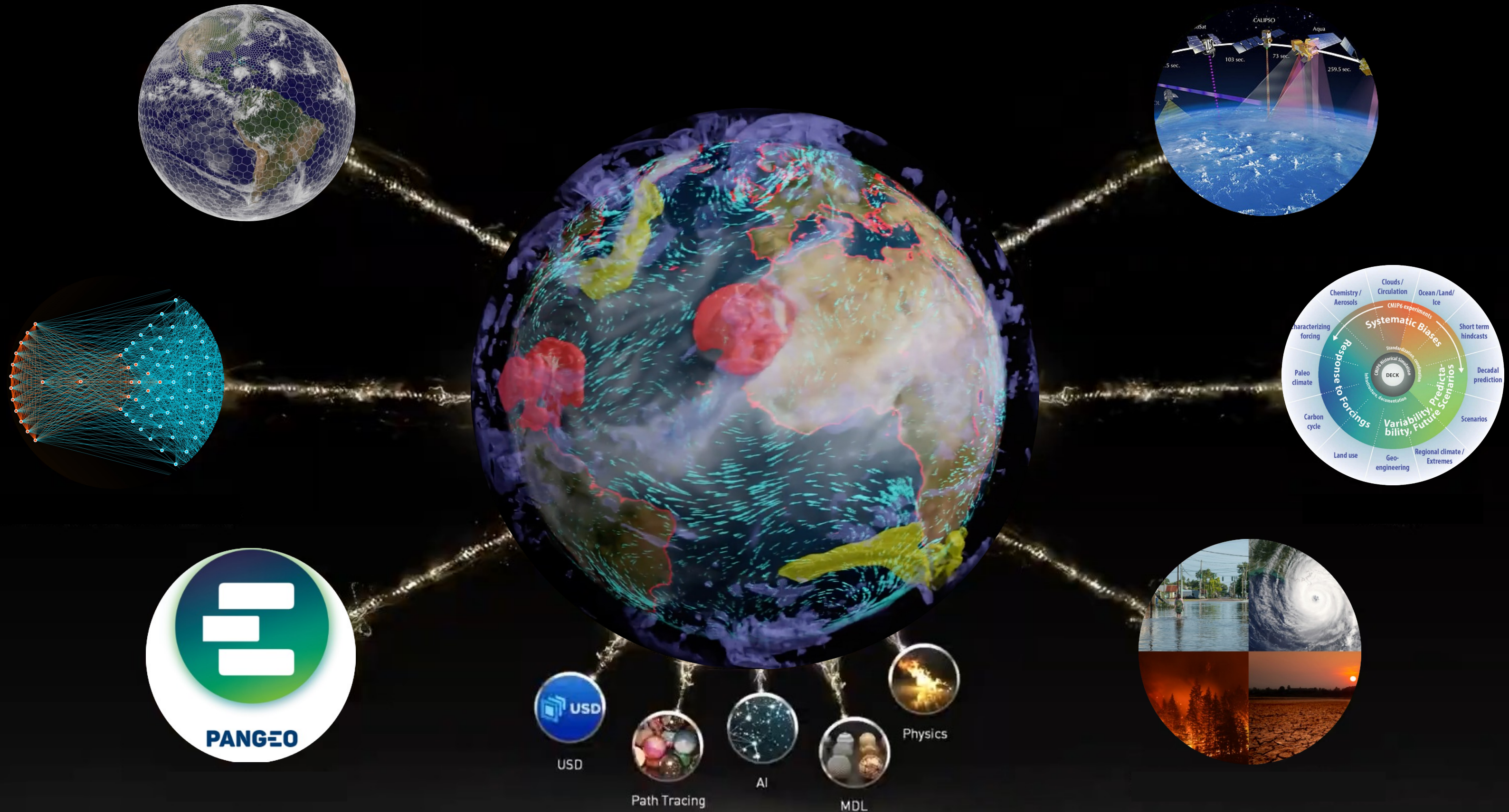
Cloud-native end-to-end MLOps pipeline to push frontiers of AI-driven weather and climate research

- End-to-end MLOps pipeline for data ingest, processing, training, inference, deployment
- NVIDIA Super-Cloud hosted
- API for easy access to popular data sources: models, reanalyses, and observations
- Leveraging Modulus Physics-ML framework and NVIDIA AI and performance optimization tools
- Fast inference and deployment
- Pre-trained models and transfer learning
- Recipes for model development and fine-tuning for regional prediction, specific phenomena (cyclones, heat waves, etc.)



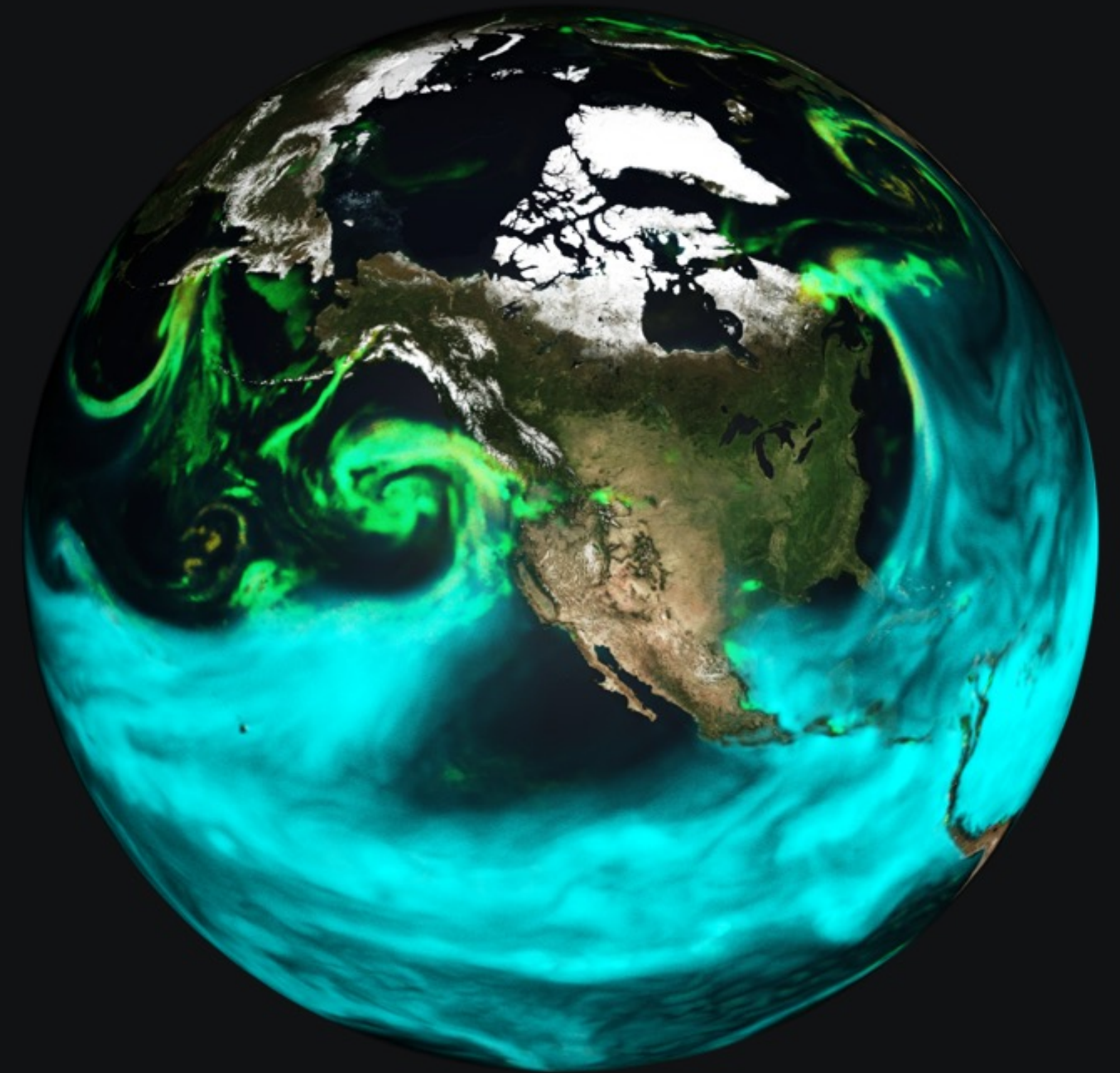
# Omniverse will enable scientists to create digital twins together

Nucleus: A shared space where models, data, tools, services, and applications synchronize



# The Vision of Earth-2

Is Beginning to Take Shape



Acknowledging: Mike Pritchard, Anima Anandkumar, David Hall, Jaideep Pathak, Noah Brenowitz, Yair Cohen, Thorsten Kurth, Boris Bonev, Christian Hundt, Andre Graubner, Peter Messmer, Stan Posey, Akshay Subramaniam, Sanjay Choudhry, Farah Hariri, Niklas Roebler, Ram Cherukuri, Nicholas Geneva, Mathias Hummel, Christopher Lamb, Mike Houston, Kamyar Azizzadenesheli, Jean Kossaifi, Steffen Roemer, Marius Koch & David Appelhans, many more NV staff & our **generous external climate science advisors Bjorn Stevens, Peter Deuben, Peter Bauer, Nils Wedi, Thomas Schulthess, and Francisco Doblas-Reyes.**