

Predicting Extreme Weather Events

Augmenting Al Models to Improve Reliability

PyData Southampton

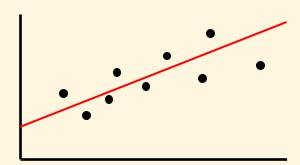
21st October 2025



What is a Machine Learning Model?

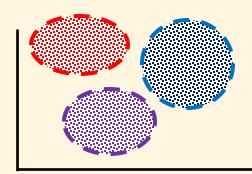
ML Model: A program created by training an algorithm on a dataset to find patterns and make predictions or decisions about new, unseen data.

Supervised Learning



Learns from labelled data

Unsupervised Learning

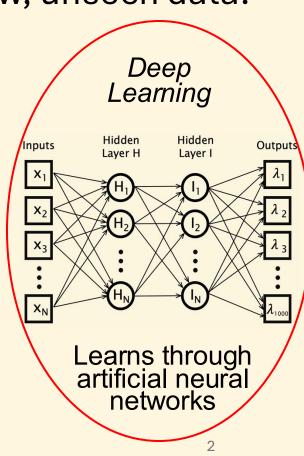


Learns from unlabeled data

Reinforcement Learning

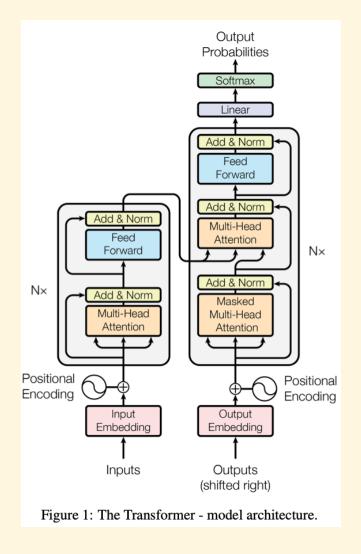


Learns through trial and error/rewards



THE FUTURE

Training Models



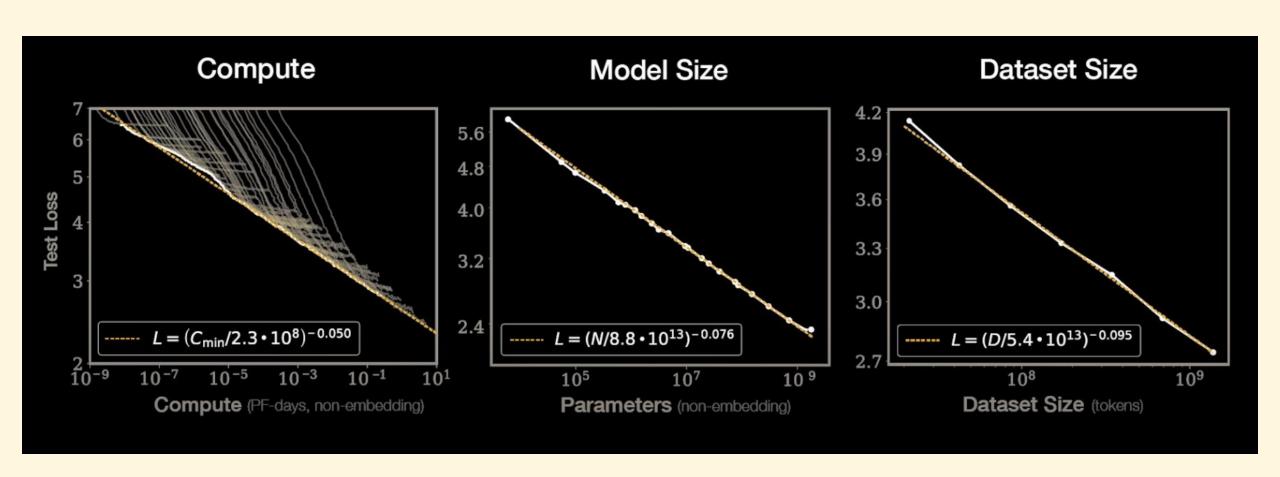
There are many deep learning architectures:

- Multi-Layer perceptrons
- Convolutional Neural Networks
- Recurrent Neural Networks
- Long Short-Term Memory
- Generative Adversarial Networks
- Autoencoders
 - •



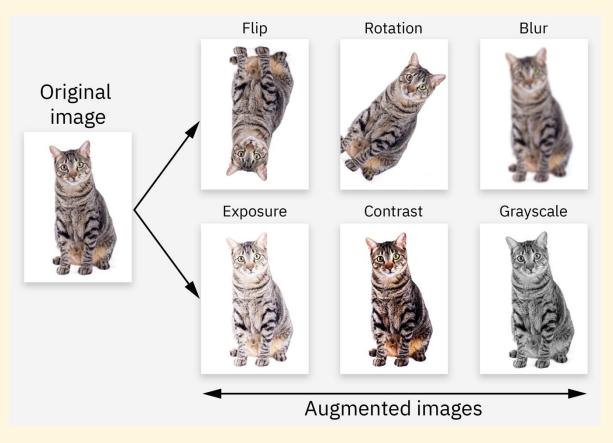
Today's most talked-about architecture is <u>Transformers</u> ... and the larger, the better.

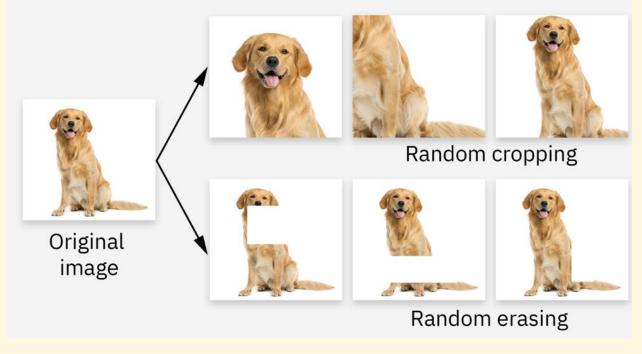
Transformer Scaling Laws



Adapting Data for Better Models

Data augmentation: uses pre-existing data to create new data samples that can improve model optimisation and generalizability.

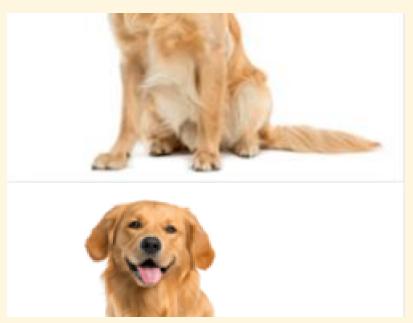


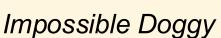


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Noise Risks

Scaling datasets poses risks and noise into the models! For example:



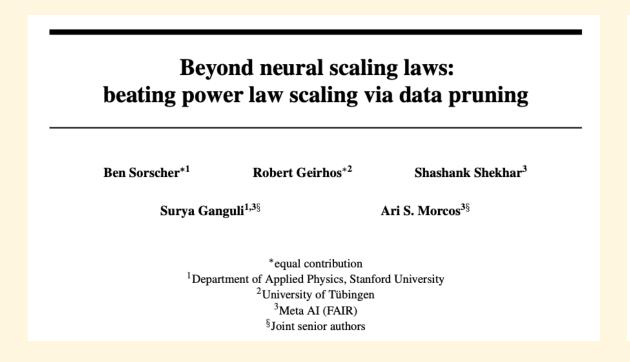


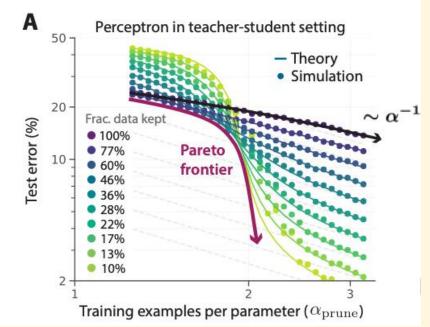


Pruning Data

Equally, not all data is equal!

Selective pruning can beat neural scaling laws.



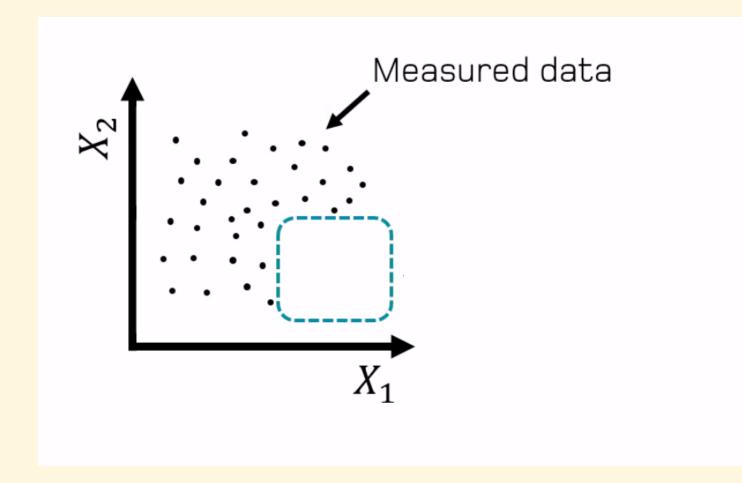


Improving Results for Science

This works great for existing datasets.

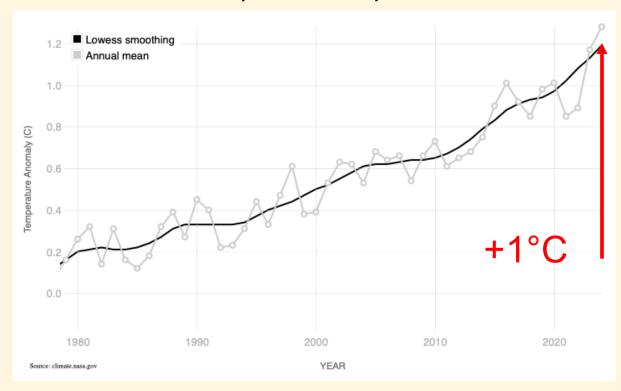
However, what happens if your data doesn't exist?

For example, how do extreme weather events react with warmer global temperatures?

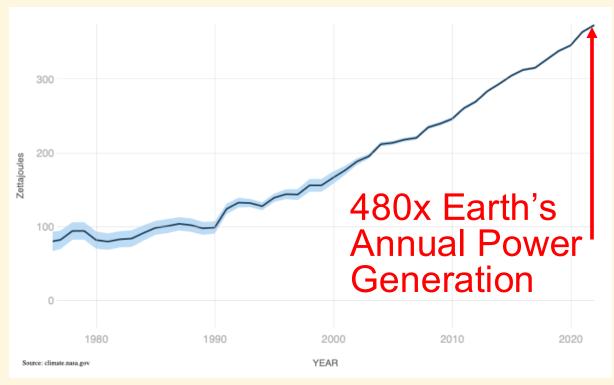


Energy absorbed by Atmosphere and Oceans

Atmospheric Temperature



Ocean Temperature



From 1980 to Today

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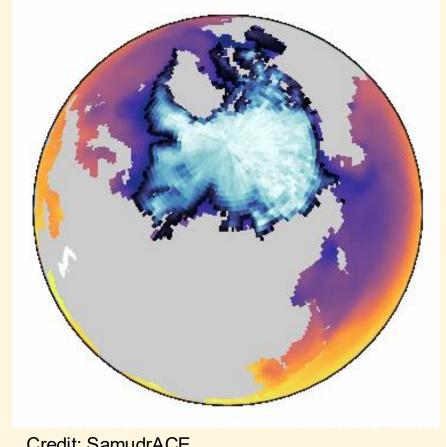
Albedo, Ice Melt and CO₂

Ice melt is decreasing Earth's albedo

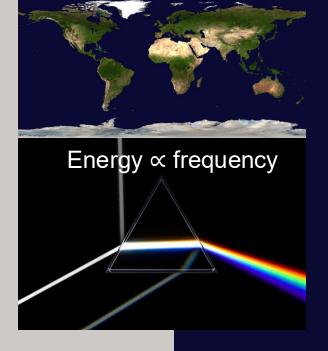
$$Albedo = \frac{Reflected\ Light}{Absorbed\ Light}$$

- Albedo change equivalent to a quarter of CO₂ emission in last 30yrs (Pistone et al. 2014)
- Expected ice-free summers in the next 70 yrs

Sea temperatures and ice coverage - 40-year rollout



Credit: SamudrACE



Ice Colour

Ocean Colour

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Out-of-distribution events

New Records Yearly!

How do we model and predict these events?

NEWS / HURRICANE

Erick, the strongest hurricane to hit Mexico so early in the year, triggers deadly flooding

By Mary Gilbert, CNN

20 Jun, 2025 2:57 PM BST | Updated 20 Jun, 2025 2:57 PM BST

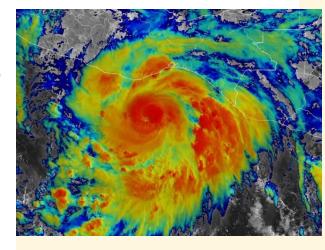
TROPICAL CYCLONES

Record-setting Atlantic storm Andrea dissipates within 12 hours of forming

By Rishav Kothari • Wednesday, June 25, 2025 •

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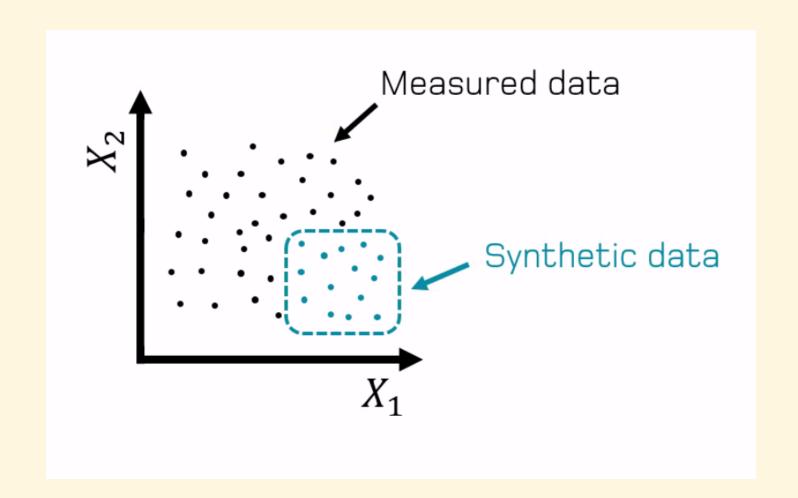
Tropical Storm Andrea weakened into a post-tropical cyclone less than 12 hours after formation as the first named storm of the 2025 Atlantic hurricane season on June 24, 2025, and is forecast to soon dissipate on June 25. Andrea has set a new record as the farthest north and east a tropical system has formed in the Atlantic during June since 1851.



Blending Physical and ML Modelling Methods

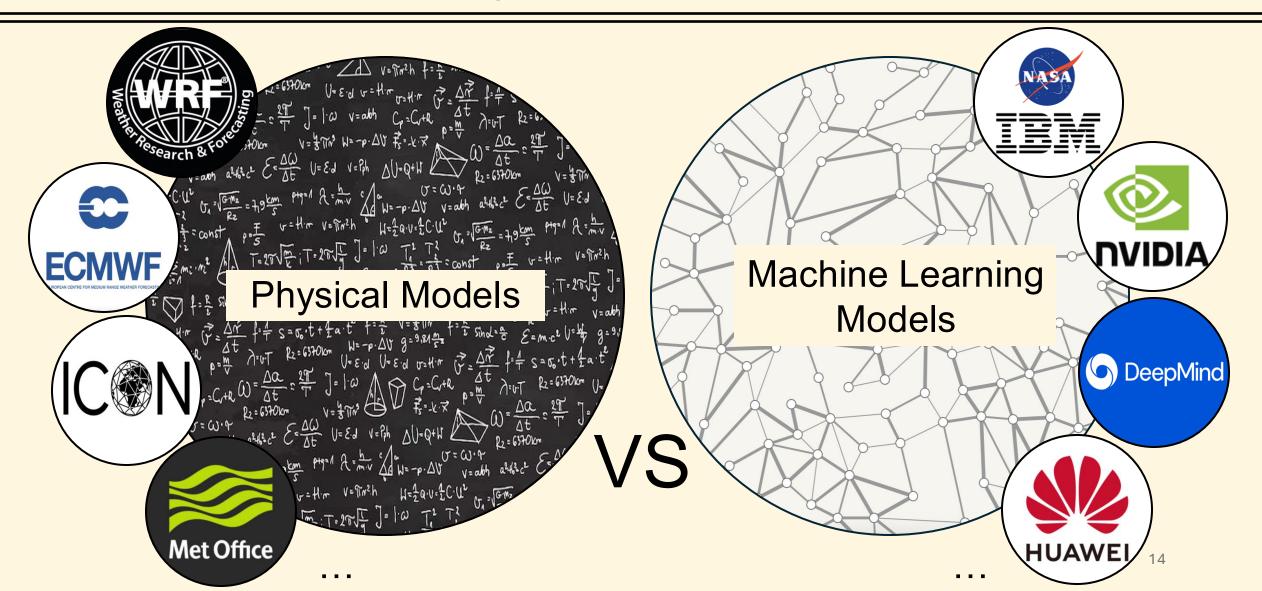
 We want to create synthetic data that embodies the underlying physics to add observation bias into historic datasets.

 This leads to robuster ML models.



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Weather Modelling Toolkits



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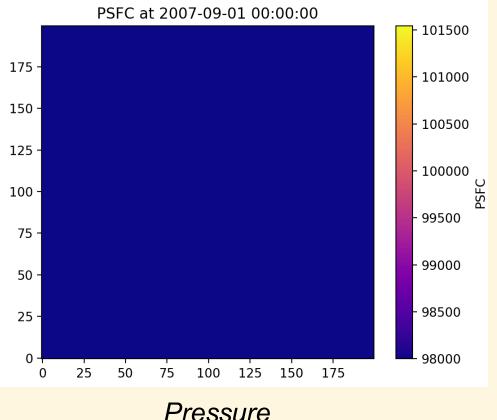
Physical Modelling - WRF

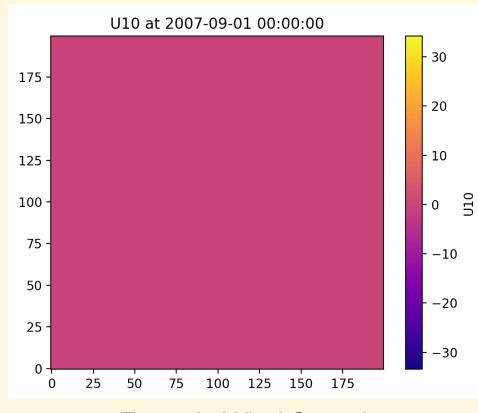
The weather research and forecasting (WRF) tool is a deterministic, physics-based forecasting tool to predict bounded weather phenomena.











Easterly Wind Speed

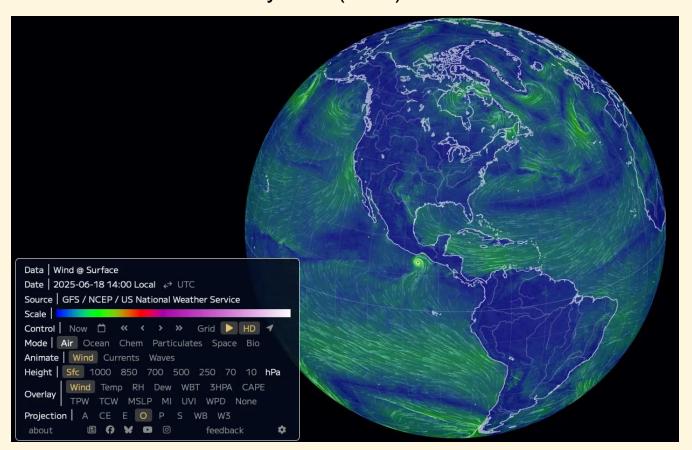
Machine Learning Models

Program	Company	Architecture
GraphCast	Google DeepMind	Graph Neural Network
GenCast	Google DeepMind	Diffusion Model
FourCastNet v3	OVIDIA	Adaptive Fourier Neural Operators
Pangu-Weather	HUAWEI	3D Earth-specific transformer
Aurora AI	Microsoft	Autoencoder/Transformer
Prithvi	IBW N	Temporal Vision Transformer

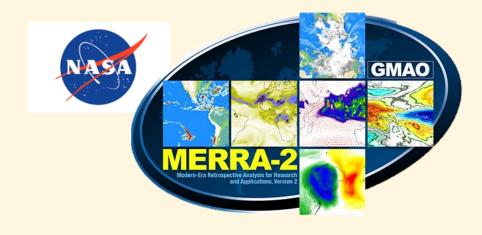
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Data Sources

Global Forecast System (GFS) Data i.e. Live Data



Historic Data - From 1940 to Today





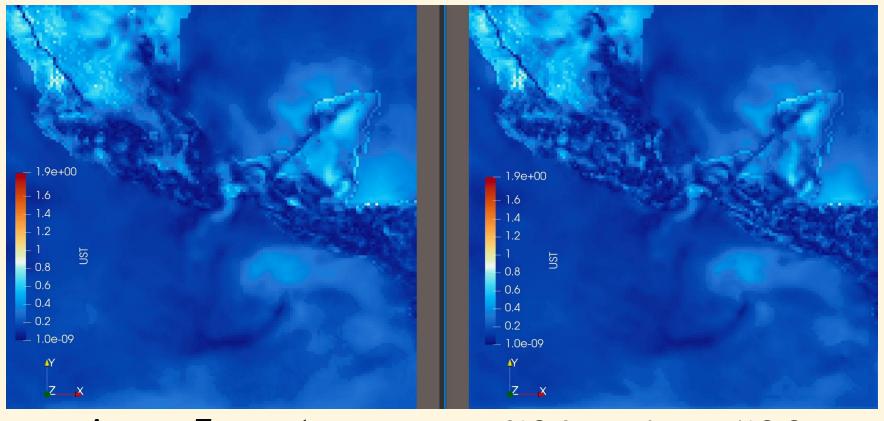
ECMWF Reanalysis v5 (ERA5)

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earth.nullschool.net

Running WRF Models

Model different parameter scenarios but not impossible scenarios

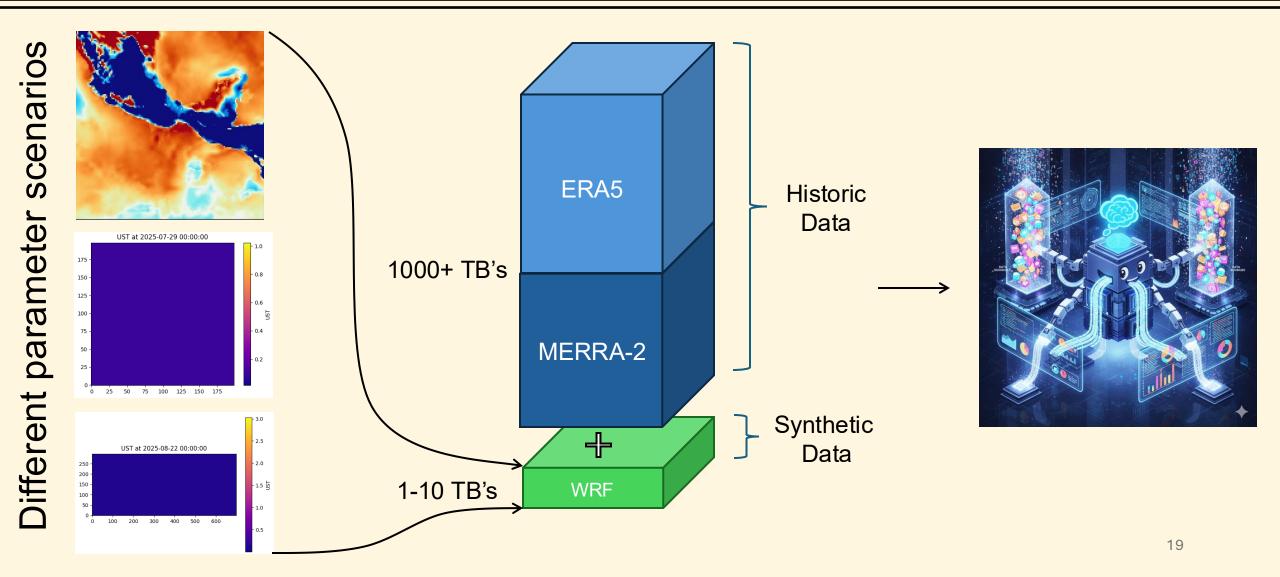


Average Temperatures

tures +2°C Atmosphere, +1°C Ocean (UST = Friction Velocity)

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Generate High Volume of Synthetic Data



Merge or Fine-tune ML Model

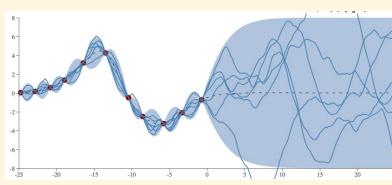
Different parameter scenarios

Temperature The Merra-2 Data Cube ACHINE LEARNING CLIMATE DATA WEATHER MODELS AUGMENTING MODELS THE FUTURE

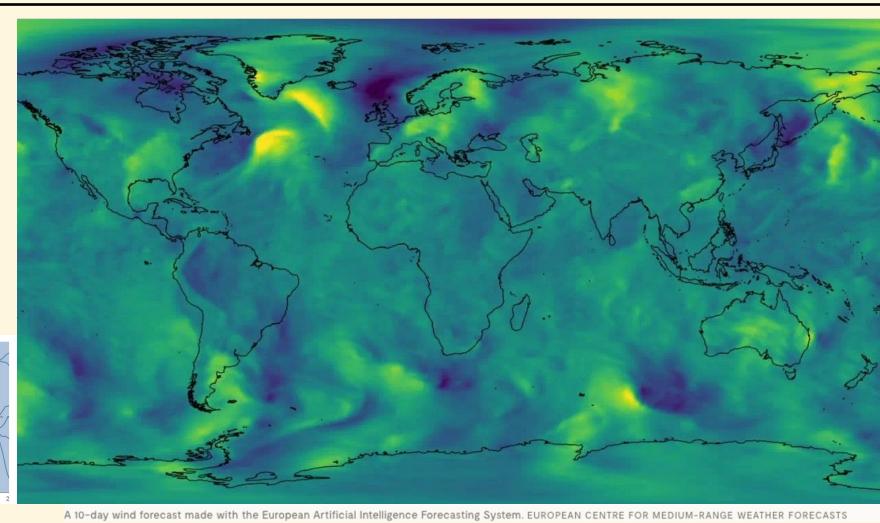
Improving Mid-Range Forecasts

Extrapolation is hard!

True representation lost at longer rollouts, so mean of ensembles used.

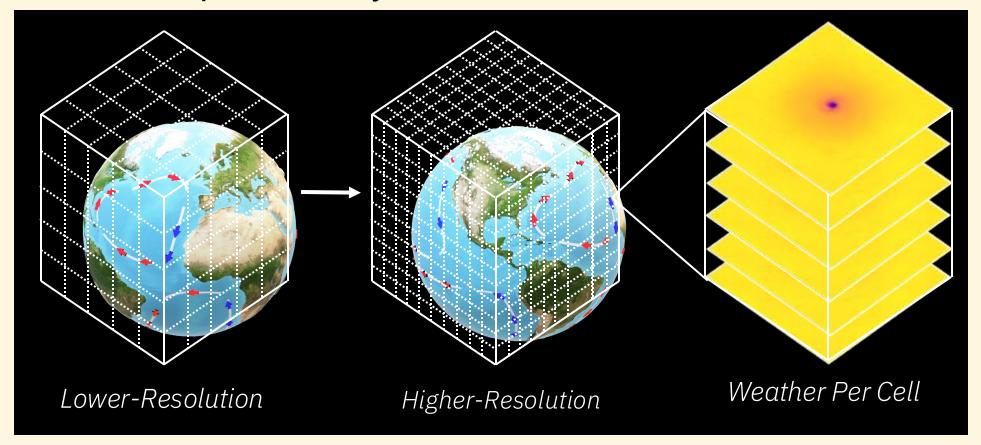


A Gaussian Process Extrapolating



High Fidelity Resolutions

- A more chaotic nature at smaller distances
- Much more computationally intensive



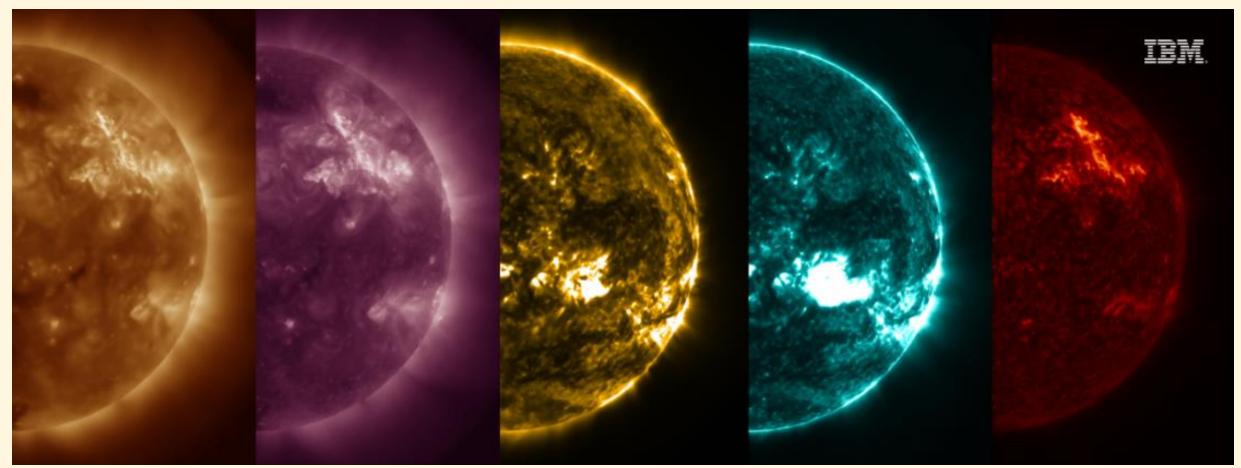
Benchmarking Unseen Events

How



Solar Flare Predictions

Space weather predictions with IBM's Suyra model





Abstract

Machine learning thrives on data, and the common wisdom is simple: the more, the better. A widely used approach to expand training datasets is data augmentation, which can enhance model robustness. Yet, augmentation is not without risk—it can also degrade performance by introducing noise. More critically, there are cases where the required training data simply does not exist, especially when models are applied to previously unseen regimes.

Nowhere is this more relevant than in the context of improving weather model predictions for extreme events, such as hurricanes. Climate change is driving our atmosphere into uncharted territory, producing extremes absent from recorded history. How, then, can we trust model predictions under such conditions? One promising avenue is through the generation of synthetic data. Hence, in this talk, we will explore how weather data can be simulated, the ML model architectures we use that are designed to forecast extreme events, and the growing competition to build the most powerful foundational weather model.