## **Articles**

- 1. Andersson, T.R., Hosking, J.S., Pérez-Ortiz, M. et al. Seasonal Arctic sea ice forecasting with probabilistic deep learning. Nat Commun 12, 5124 (2021). https://doi.org/10.1038/s41467-021-25257-4
- 2. Price, I., Sanchez-Gonzalez, A., Alet, F. et al. Probabilistic weather forecasting with machine learning. Nature 637, 84–90 (2025). <a href="https://doi.org/10.1038/s41586-024-08252-9">https://doi.org/10.1038/s41586-024-08252-9</a>
- 3. Thomas Vandal, Evan Kodra, Sangram Ganguly, Andrew Michaelis, Ramakrishna Nemani, and Auroop R. Ganguly. 2017. DeepSD: Generating High Resolution Climate Change Projections through Single Image Super-Resolution. In Proceedings of the 23rd ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD '17). Association for Computing Machinery, New York, NY, USA, 1663–1672. https://doi.org/10.1145/3097983.3098004
- 4. Evan Racah, Christopher Beckham, Tegan Maharaj, Samira Ebrahimi Kahou, Prabhat, and Christopher Pal. 2017. Extreme weather: a large-scale climate dataset for semi-supervised detection, localization, and understanding of extreme weather events. In Proceedings of the 31st International Conference on Neural Information Processing Systems (NIPS'17). Curran Associates Inc., Red Hook, NY, USA, 3405–3416.
- 5. Gordon, E. M., & Barnes, E. A. (2022). Incorporating uncertainty into a regression neural network enables identification of decadal state-dependent predictability in CESM2. Geophysical Research Letters, 49, e2022GL098635.

Title: Authors: Year:

- 1. What is the ML application? [Slide 3]
- 2. What type of problem is it (i.e. classification or regression)?

What type of model is this (i.e. NN, CNN, LSTM etc.)?

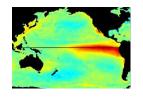
a. Bonus: Why do you think they chose this model type?

- 3. What are the inputs and outputs? What is their structure (i.e. maps, timeseries, data points)? What are the data sources?
- 4. What is the timescale (i.e. △t between inputs and outputs)?
- 5. What type of challenges and solutions were faced or applied? [Slide 4]

# ML Applications (Q1)

#### Sources of Predictability

Climate Modes Causal analysis



#### **Feature Detection**

Extremes





#### **Forecasting**

Short-term weather forecasting



Extreme Weather Events

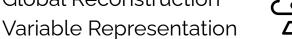
Climate Forecasting



#### **Satellite Data & Observations**

Interpolation

Global Reconstruction





Downscaling

Merging Satellite Data

Synthetic data

Sensor placement



### Earth System Modeling



Model Calibration and

Validation

**Equation Discovery** 

**Uncertainty Quantification** 

Sources of uncertainty

Parameterizations

**Bias Correction** 

Data Assimilation

Climate Model analysis and

benchmarking

# **Challenges and Solutions (Q5)**

Challenges	Solutions
Data Availability	Climate model output
Robustness on out-of-distribution samples	Transfer learning
Interpretability	eXplainable AI
Physical Inconsistency	Physics-informed ML
<b>Uncertainty Quantification</b>	Bayesian NNs, Problem Set-Up

Title:

Authors:

Type of problem:

ML application:

