# **Graph-based Neural Weather Prediction for Limited Area Modeling** Joel Oskarsson<sup>1</sup> Tomas Landelius<sup>2</sup> Fredrik Lindsten<sup>1</sup>

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Ground Truth Hi-LAM GC-LAM **Figure:** Forecasts of solar radiation (nlwrs) at 57 h lead time

## Machine Learning for Weather Forecasting

The latest Neural Weather Prediction (NeurWP) models are as accurate as traditional Numerical Weather Prediction (NWP) systems, but require far less resources.

#### Traditional NWP NeurWP



Figure: RMSE at different lead times

## Neural Limited Area Models

#### Graph-based NeurWP [1, 2]

- Construct mesh graph covering forecasting area
- Utilize Graph Neural Networks (GNNs) to
  - Encode gridded observations to mesh

Time to produce forecast	Hours	Seconds
Computational resources	Large cluster	Single GPU

**Problem formulation:** Given initial condition  $X^0$  predict a trajectory  $X^1, \ldots, X^T$  of following weather states.

- Learn single time step dynamics model  $\hat{f}$
- Roll out forecast auto-regressively



Existing models focus on global weather forecasting. Our work: Adapt graph-based neural weather prediction for Limited Area Models (LAMs)!

## Climate Impact

- Enabling large-scale ensemble forecasting, important for extreme weather prediction
- Providing fast and accurate forecasts for renewable en-

- Process latent representation on mesh
- Decode to grid, producing one time step prediction
- GC-LAM: adaptation of GraphCast [2] to LAM
- Hi-LAM: Extended model with hierarchical mesh graph



## Adapting to LAM setting

- LAM mesh graph: Regular quadrilateral mesh
- Area boundary: Forcing along boundary from existing forecast



#### ergy production

Reducing the energy required to produce forecasts

# Dataset: MEPS forecasts

- 6000 forecasts from the MetCoOp Ensemble Prediction System (MEPS)
- Data in  $238 \times 268$  grid, covering Nordic area
- 17 atmospheric variables: Temperature, wind, pressure, water vapor, solar radiation, humidity, geopotential

## Contact and Links





We are attending virtually, please ask questions through the submission webpage or later at joel.oskarsson@liu.se.

Webpage Paper

## References

- [1] R. Keisler. Forecasting global weather with graph neural networks. *arXiv preprint arXiv:2202.07575*, 2022.
- [2] R. Lam et al. Learning skillful medium-range global weather forecasting. *Science*, 2023.

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