

# Towards ensemble data assimilation of satellite radiances in Machine Learning Limited Area Models



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## INTRODUCTION

### 1) Need to exploit satellite radiances better in convective-scale assimilation

Satellite radiances assimilation is crucial for Numerical Weather Predictions (NWP) in global models. However, radiances are under-used in Limited Area Models (LAM), especially at microwaves.

### 2) Machine Learning Weather Predictions (MLWP) complementing or replacing NWP

Several global MLWP models exist, but with limited resolution. Recent attempts to build high-resolution Machine Learning Limited Area Models (ML LAM) up to convective-scale<sup>[1]</sup>.

Based on these aspects, we first show the implementation of **ensemble-based assimilation of microwave and infrared water vapour channels** in a operational NWP LAM.

- Microwaves: low spatio-temporal resolution, good cloud penetration.
- Infrared: high spatio-temporal resolution, no cloud penetration.

This motivates the development towards radiances assimilation in a ML LAM. Methodology, preliminary results, main scientific challenges, and future outlooks are described.

## 1. RADIANCES ASSIMILATION IN NWP LAM

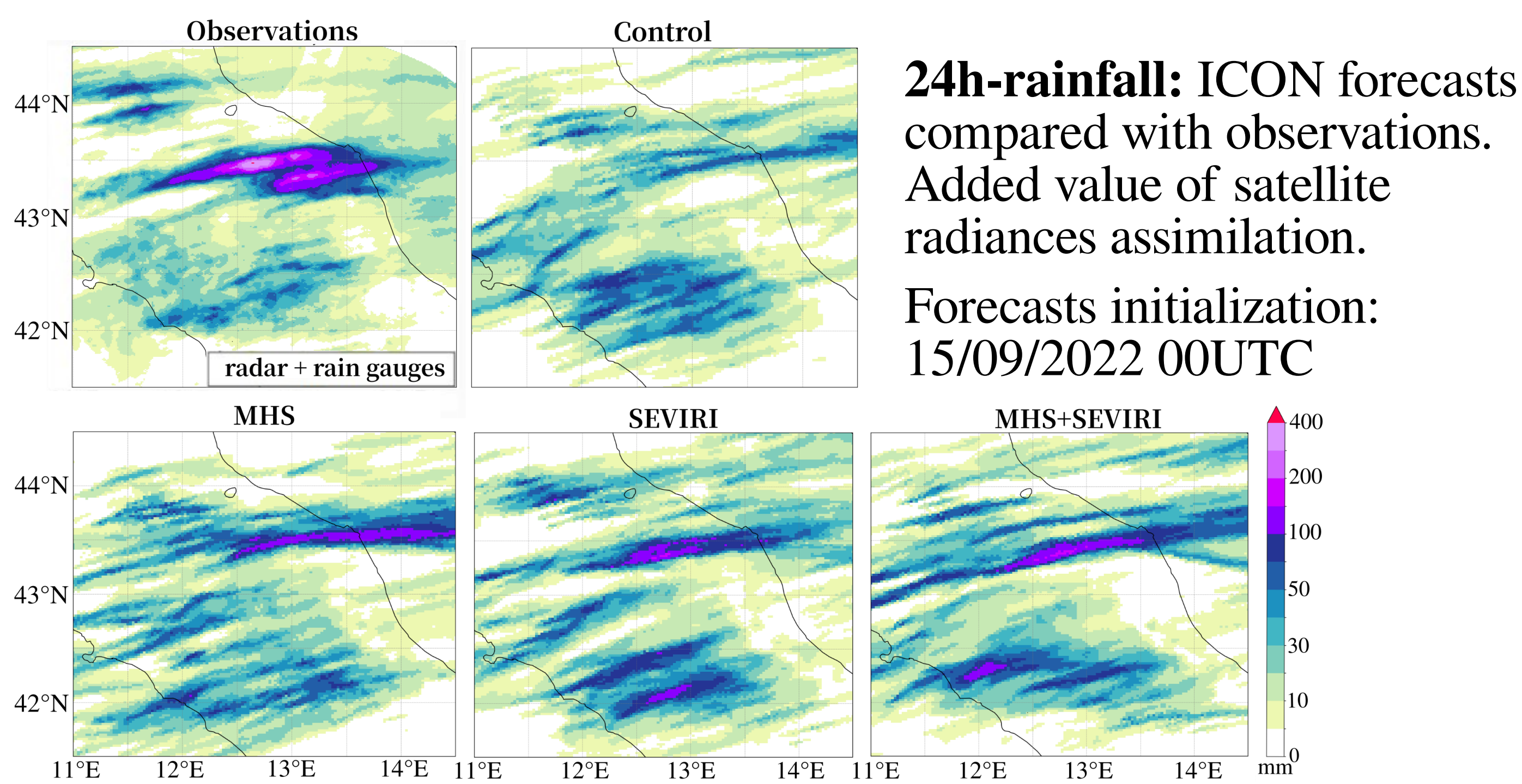
ICON: *ICO*sahedral Non-hydrostatic model at 2.2km resolution

KENDA: *Kilometre-Scale Ensemble Data Assimilation* system, based on a Local Ensemble Transform Kalman Filter (LETKF), 40 members, 1h-cycle

Severe convection case study in Marche, Central Italy (12-16 Sept. 2022)<sup>[2]</sup>

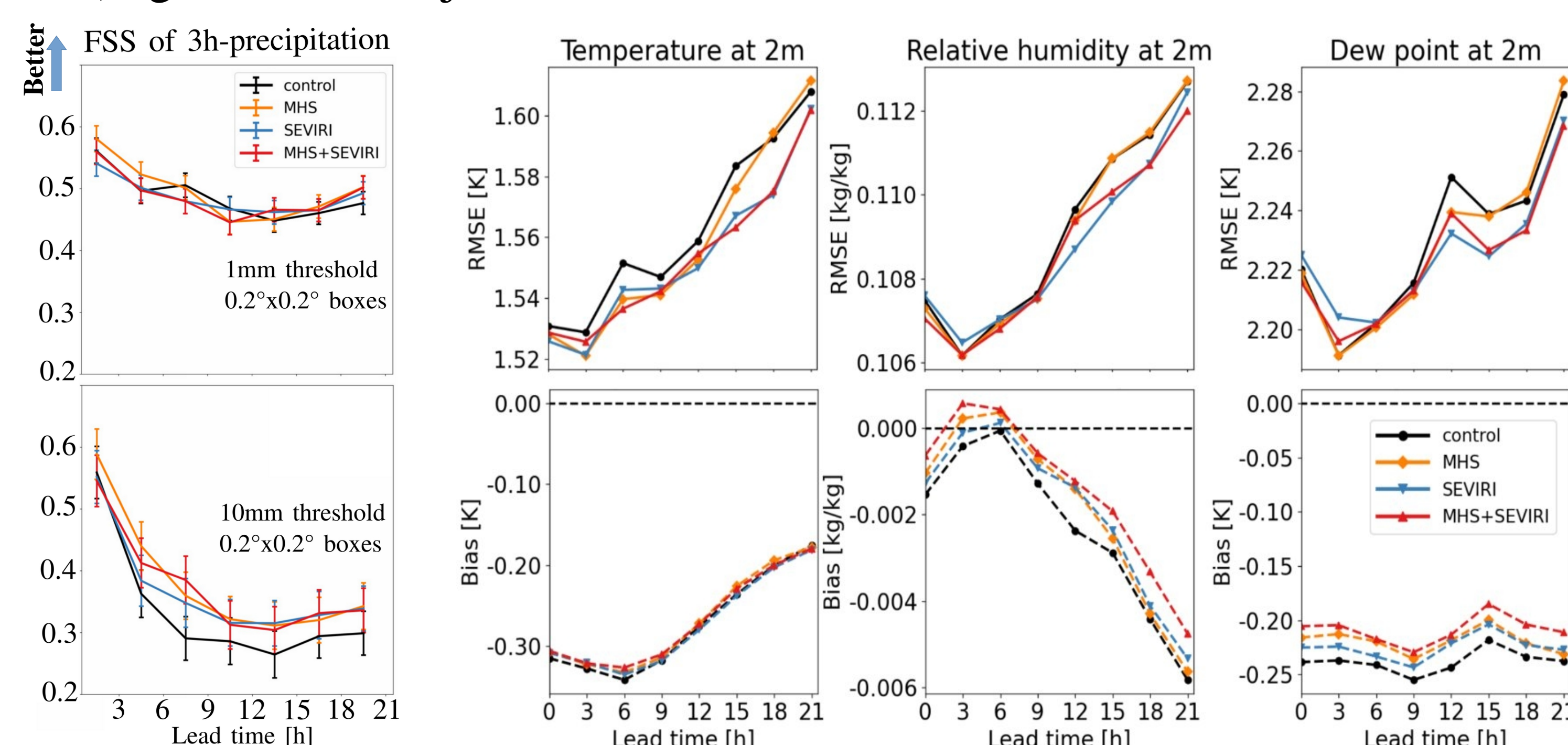
Assimilated observations	control	MHS	SEVIRI	MHS+SEVIRI
Conventional (SYNOP, AIREP, TEMP)	✓	✓	✓	✓
Radar volumes (reflectivity and radial winds)	✓	✓	✓	✓
Latent Heat Nudging (LHN)	✓	✓	✓	✓
Microwave Humidity Sounder (MHS) chan. 3, 4, 5 in <i>clear-sky</i> conditions	✗	✓	✗	✓
SEVIRI chan. 5, 6 in <i>all-sky</i> conditions	✗	✗	✓	✓

Operational setup of ItaliaMeteo and ArpaE



Bottom left column: precipitation verification (Fractions Skill Score, FSS) against radar-adjusted data.

Bottom right: surface variables verification (RMSE and bias) against ground stations data.

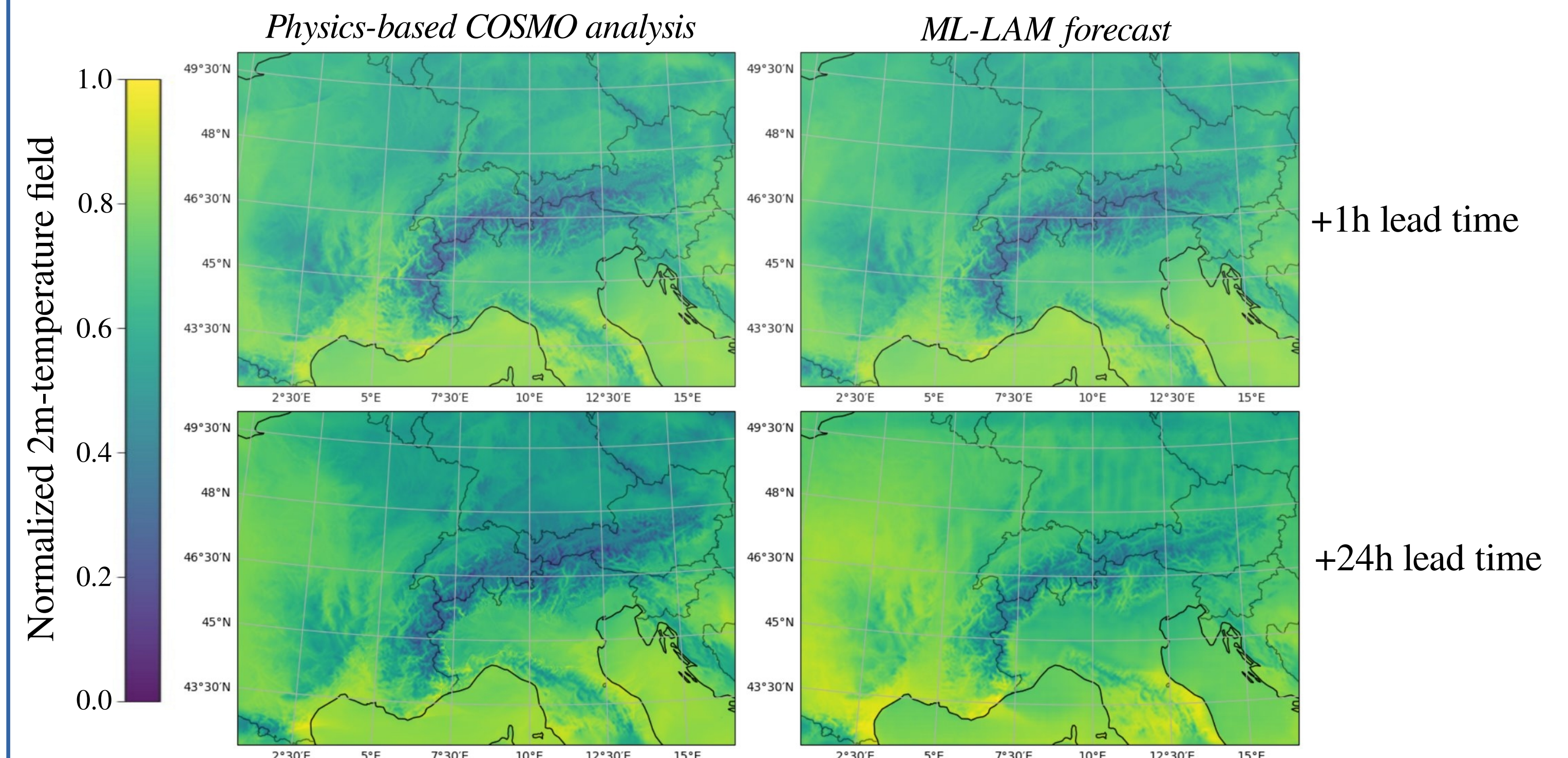


## 2. PRELIMINARY TESTS WITH ML LAM

### ML LAM setting by Adamov et al., 2025<sup>[1]</sup>

- Trained on operational analyses dataset over Switzerland generated with COSMO model and KENDA system
- GNN with encoder-processor-decoder architecture
- MSE loss function, 8 vertical levels, 2.2km resolution

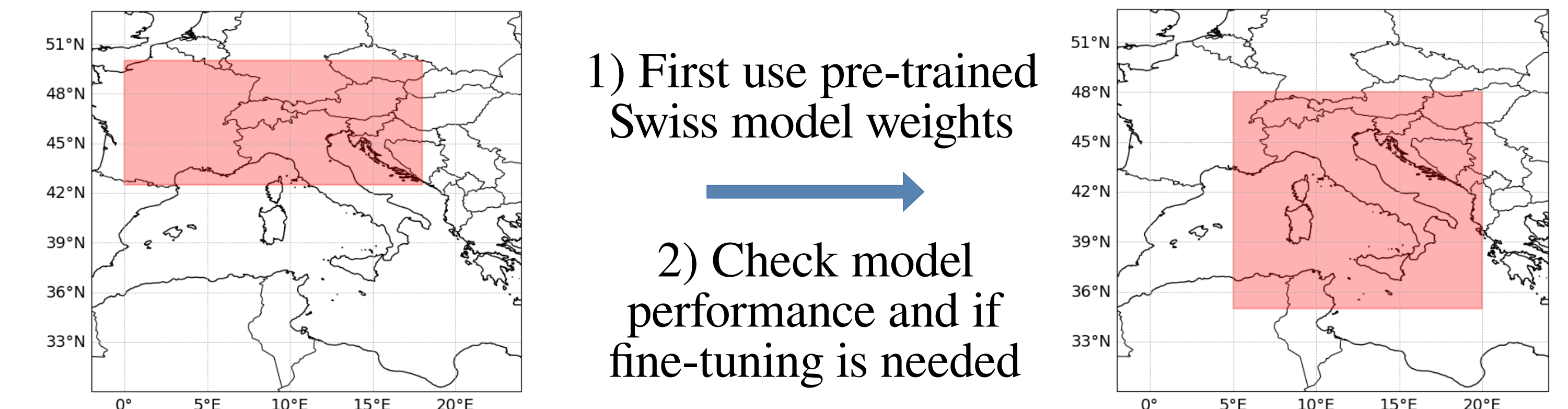
We implemented this open-source ML LAM on Leonardo supercomputer in Bologna. 24h forecast takes 4/5 minutes on A100 GPU.



Next step: running the model over Italy in severe convection conditions.

COSMO analysis data from ArpaE Emilia-Romagna used as initial conditions:

- 2018-08-25 to 2025-09-30
- Archived in GRIB1 format, need conversion to Zarr



## 3. TOWARDS RADIANCES ASSIMILATION IN ML LAM

Can we successfully perform ensemble-based assimilation of radiances using this ML LAM as forecast model?

### Challenges:

**Few model vertical levels:** detrimental for radiative transfer computations.

→ Evaluate the performance of RTTOV forward operator applied on ML LAM fields with low number of model levels.

**Under-dispersion:** missing upscale evolution of small perturbations, detrimental for ensemble forecasting.

→ Test boundary conditions perturbations to increase LETKF ensemble spread.

**Fields smoothing:** quadratic loss function training

→ Evaluate the power spectrum of ML LAM on 1h lead time forecasts (i.e. analysis cycle frequency)

## CONCLUSIONS

### Key results

- Assimilation of water vapour satellite channels: improving forecast of a severe convection event in an operational NWP LAM<sup>[3]</sup>.
- Successfully running a convective-scale ML LAM trained on Switzerland<sup>[1]</sup>.

### Current work

- Evaluate the performance of ML LAM on Italy in severe convection conditions.
- Check for possible fine-tuning over Italy.

### Future outlooks

- Address challenges towards ensemble assimilation of radiances in ML LAM.

## References

- [1] Adamov, S. et al., *Building Machine Learning Limited Area Models: Kilometer-Scale Weather Forecasting in Realistic Settings*, arXiv preprint, 2025.  
 [2] Grenzi, M., *Towards the assimilation of satellite radiances in the convection-permitting ICON model*, NCC, 2025.  
 [3] Grenzi, M. et al., *Microwave radiances assimilation for severe convection forecasting in an operational limited-area model*, QJRMS, 2026 (submitted).

## Contacts

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