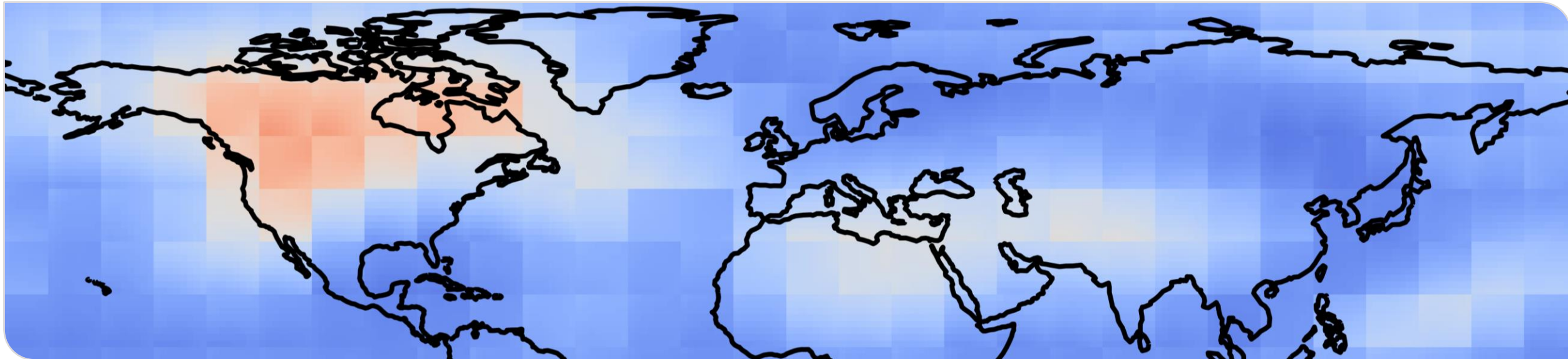
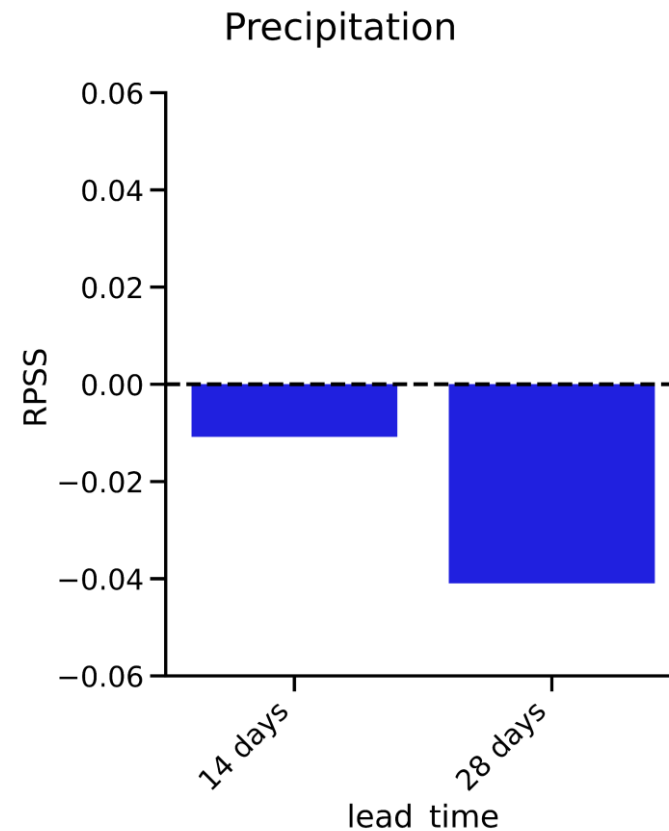
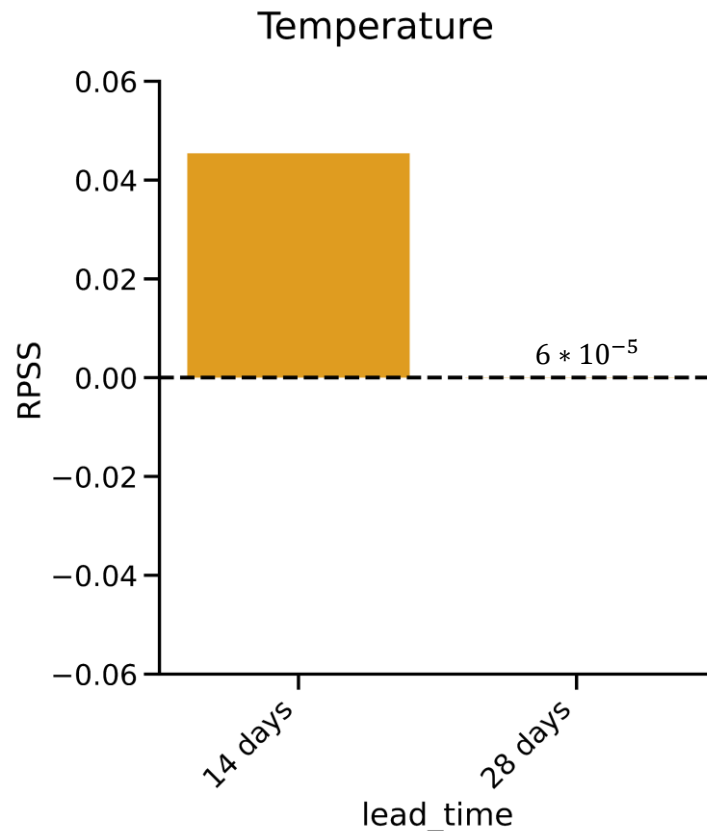


Machine Learning for global probabilistic prediction of temperature and precipitation on sub-seasonal time-scales

Nina Horat, Sebastian Lerch



Challenge to improve Sub-seasonal to Seasonal Predictions using Artificial Intelligence (2021)

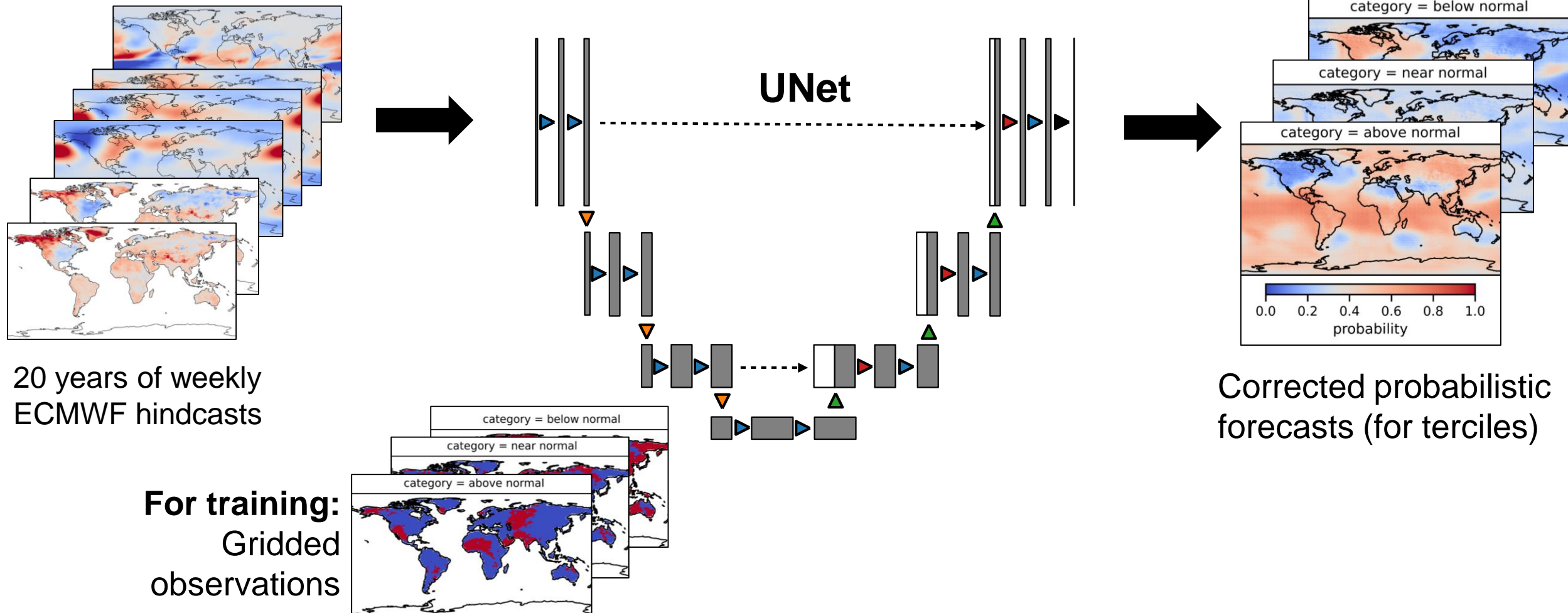


- Globally aggregated RPSS for all Thursday forecasts in 2020.
- $RPSS > 0$: forecast is better than climatology (random guessing).

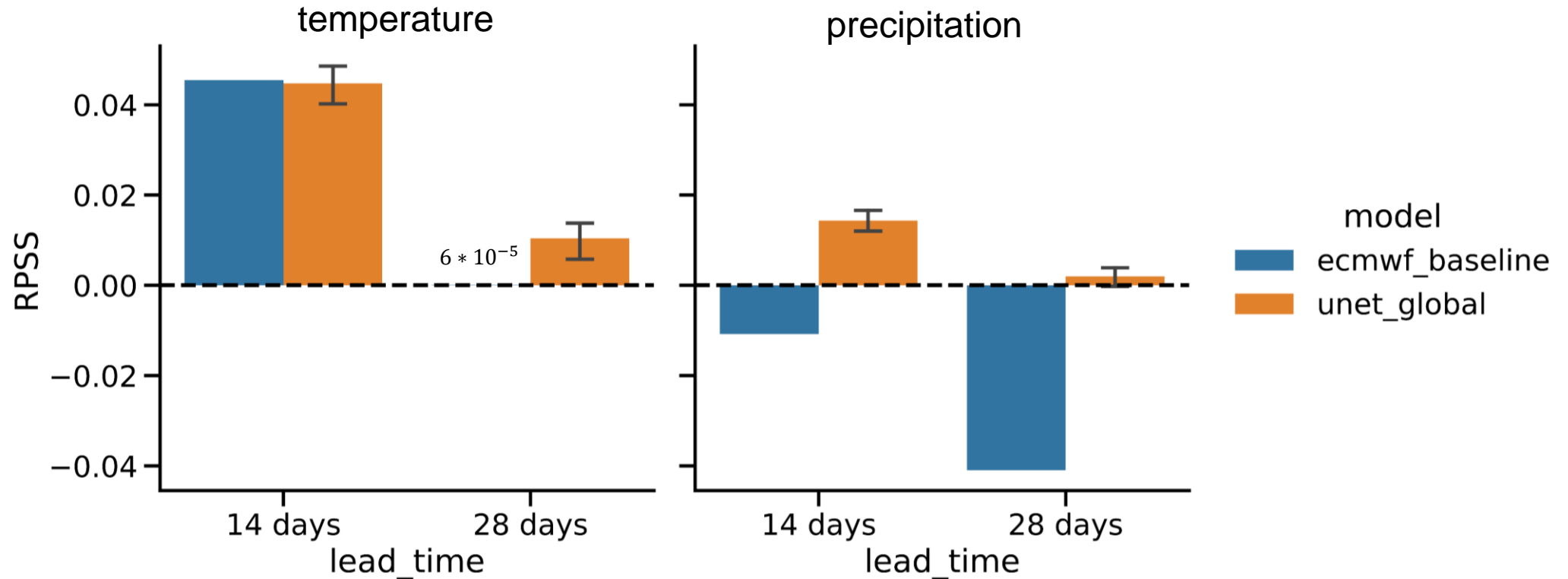
Task of the S2S AI challenge:

- Probabilistic predictions for terciles of
 - Average 2m-temperature
 - Accumulated precipitation
 - Aggregated over 2 weeks
- For 2 lead times:
 - 14 days: weeks 3-4
 - 28 days: weeks 5-6
- Spatial and temporal target:
 - Every Thursday in 2020
 - All land grid points globally
- <https://s2s-ai-challenge.github.io/>

Post-Processing with UNet architecture



Performance on test data (2020)



- Globally aggregated RPSS for all Thursday forecasts in 2020
- $RPSS > 0$: forecast is better than climatology (random guessing)
- Error bars: variability across folds (10-fold CV)

Conclusion

- UNet model is equally good or better than the calibrated probabilistic extended-range forecast of ECMWF.
 - Successfully transforms the ensemble mean forecast into a probabilistic forecast
 - Achieves positive skill where the ECMWF forecast has no skill
- Flexible post-processing method that can exploit spatial information and information from many different variables

- Feel free to contact me at: nina.horat@kit.edu