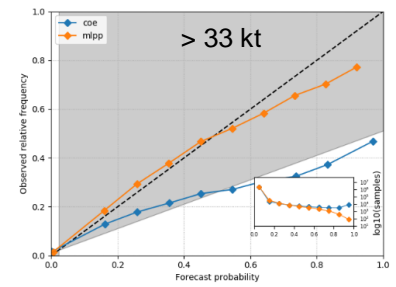
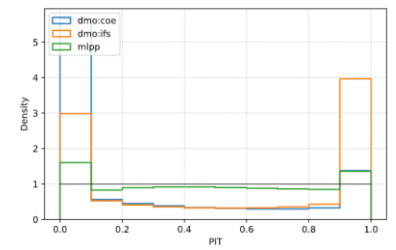
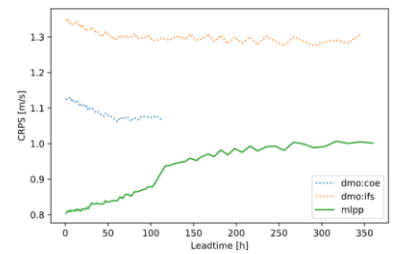
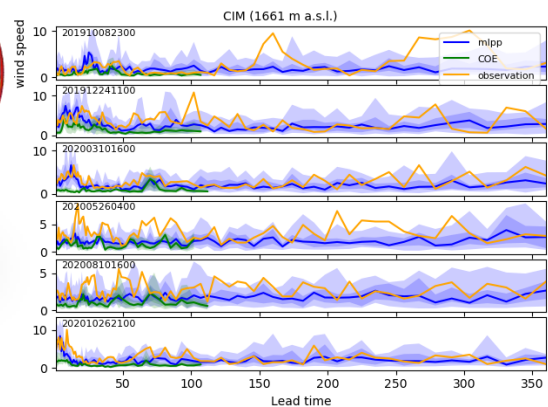
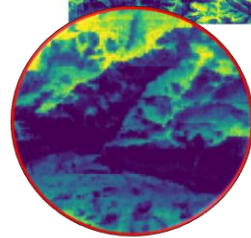
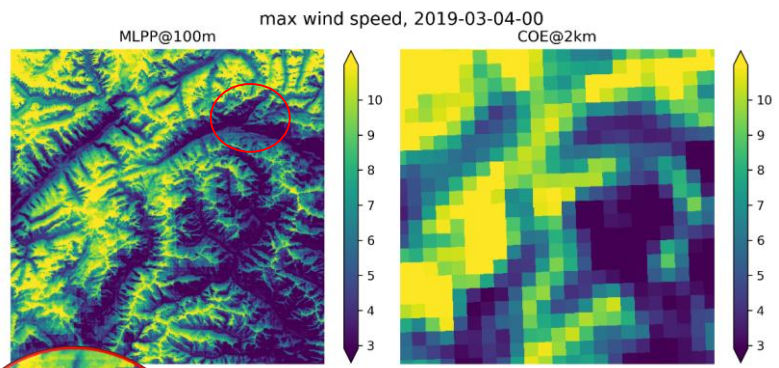
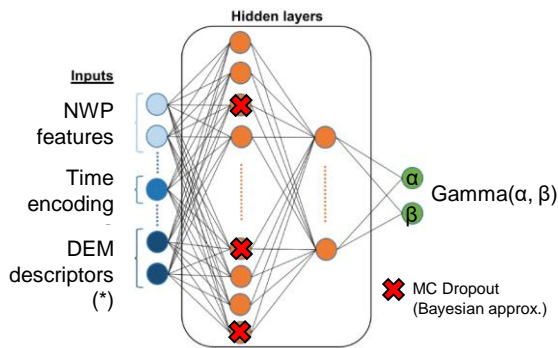


# Seamless postprocessing of multi-model NWP surface wind forecasts with deep learning



(\*) <https://github.com/MeteoSwiss/topo-descriptors>

- Aleatoric uncertainty
- Epistemic uncertainty

- NWP sources
- COSMO-1 (1km, +33h, det)
  - COSMO-E (2 km, +120h, 21 memb.)
  - IFS-ENS (18 km, +360h, 51)

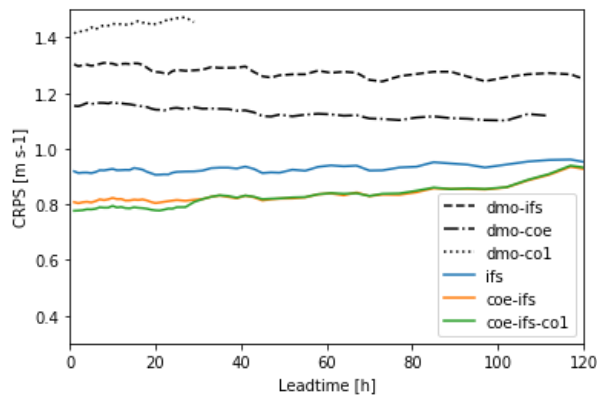
- Data split:
- train (2016-2018, 300 stations)
  - val (2019, 100 stations)
  - test (2020, 100 stations)

# Seamless postprocessing of multi-model NWP surface wind forecasts with deep learning

## Strategy for seamless

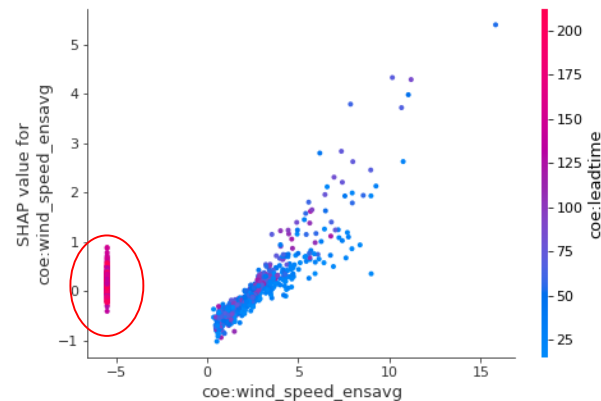
- Train one model for all NWP sources and lead times.
- Simulate model availability during training.
- Keep missing values.

→ Let the network learn from exposure to the data



## In practice:

- The NWP arrival time is a random variable  $\sim \exp(\lambda)$
- The information «missing NWP source» is imputed with a value outside of its possible range, so that the model learns to ignore it.



SHAP: <https://github.com/slundberg/shap>