



Norwegian
Meteorological
Institute

Regional Sea Ice predictions in northern Norway, Svalbard and the Barents Sea using Deep Learning

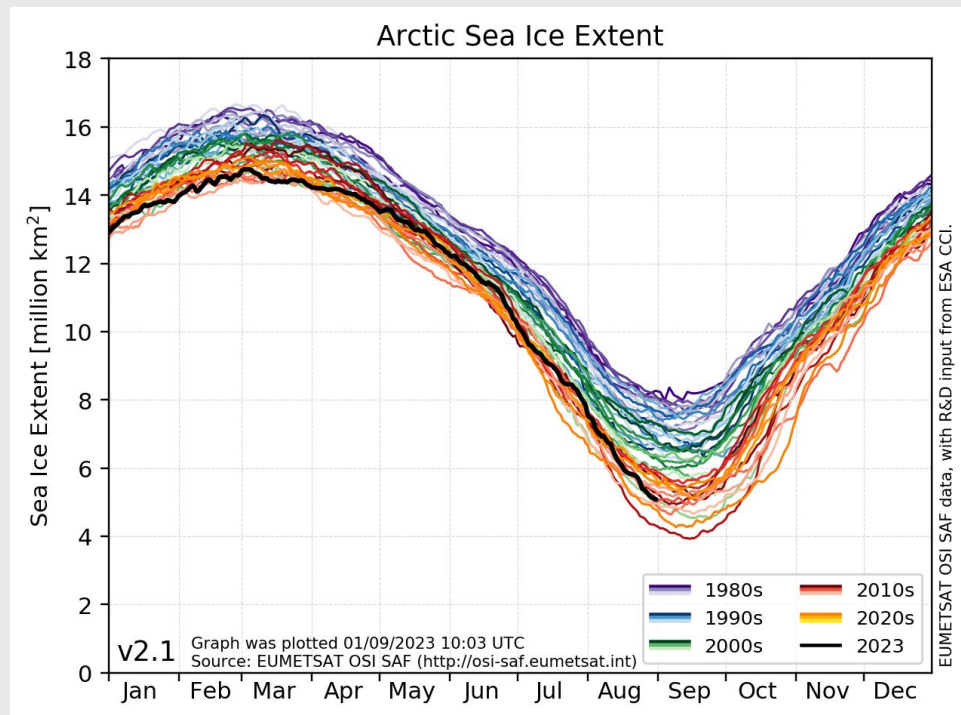
Are Frode Kvanum

04.11.2024

SuperIceWorkshop - Frascati, Italy

Problem description

- Arctic sea ice is retreating, socio-economic interest and activity in the Arctic is increasing
- Regional scale ice-condition forecasting provide valuable information to maritime operators
- Precise ice-cover information is critical for accurate Arctic weather forecasts.



Weather forecasts are sensitive to sea ice cover

- Different sea-ice products can lead to T2M variations of up to 5°C in AROME Arctic
- The atmospheric response is not only local, differences can develop hundred of kilometers away from the sea ice

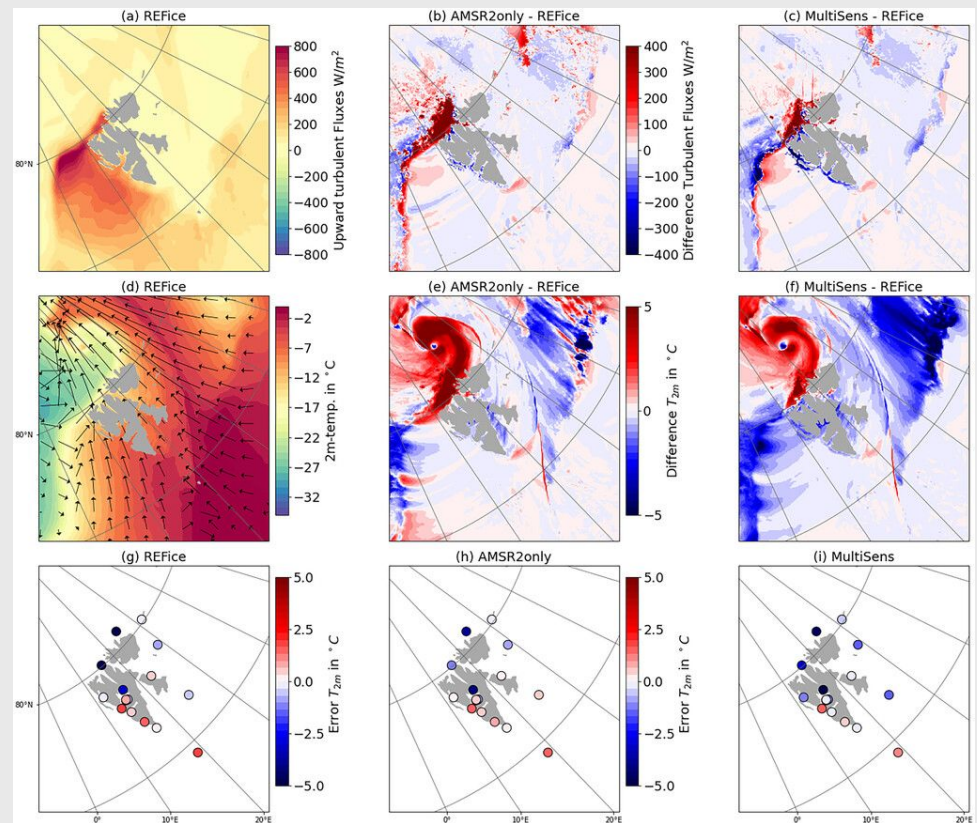


Figure from: Müller, M., Batrak, Y., Dinessen, F., Grote, R., and Wang, K.: Challenges in the Description of Sea Ice for a Kilometer-Scale Weather Forecasting System, *Weather and Forecasting*, 38, 1157–1171, <https://doi.org/10.1175/WAF-D-22-0134.1>, 2023.

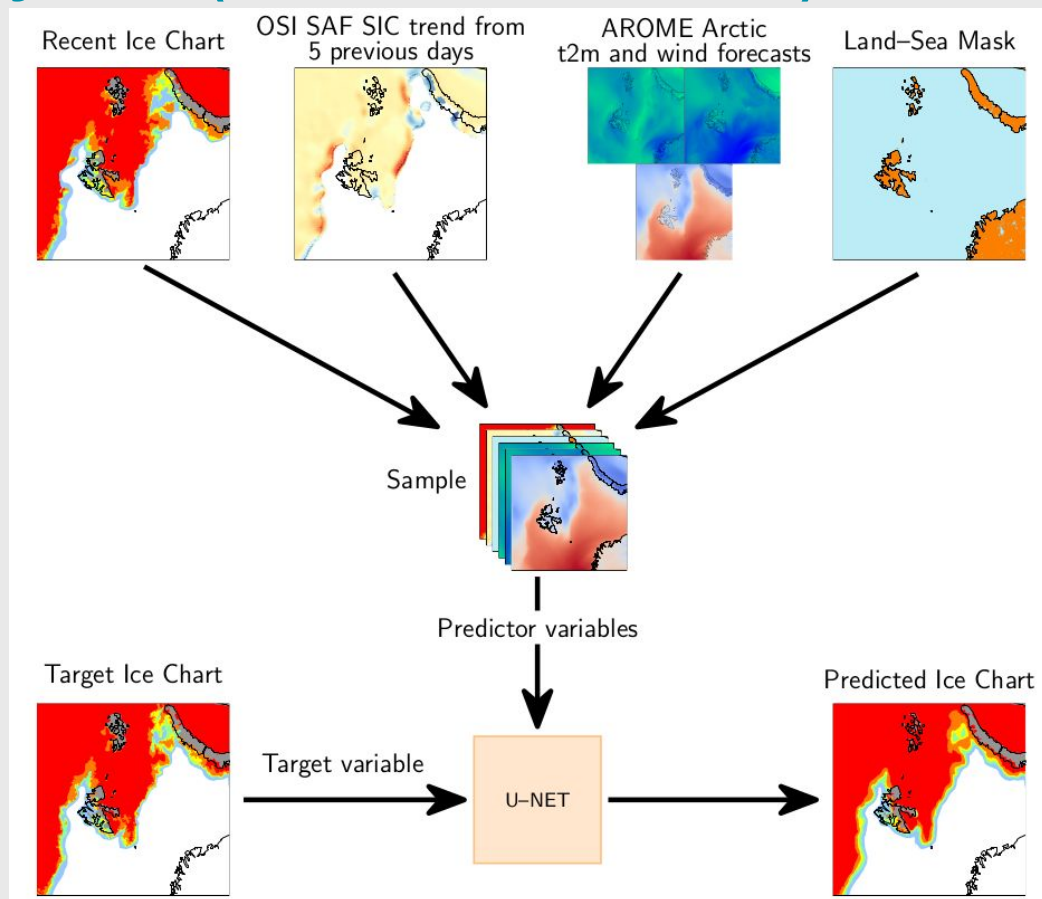
Developing a short-term (3 days) and high-resolution (1 km) regional sea ice prediction system (Kvanum et al., 2024)

❖ Predictors:

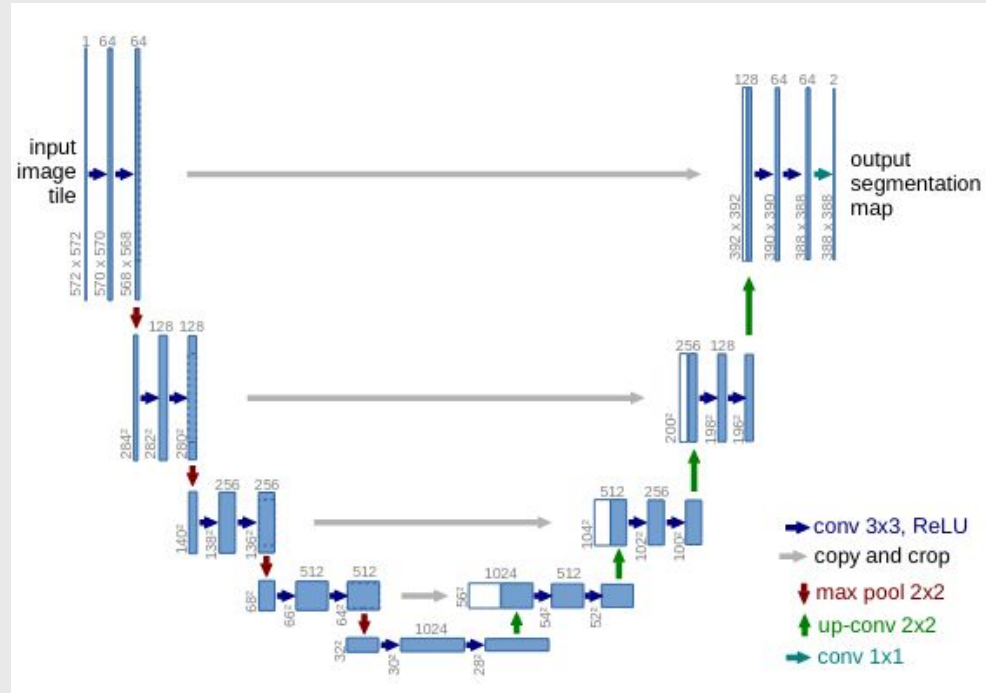
- Latest ice chart
- Sea ice concentration trend from OSI-SAF
- AROME-Arctic weather forecasts
- Land-Sea mask

❖ Target variable:

- Ice charts from the Norwegian Ice Service



Deep learning model implementation, U-Net



- Predicting cumulative contours

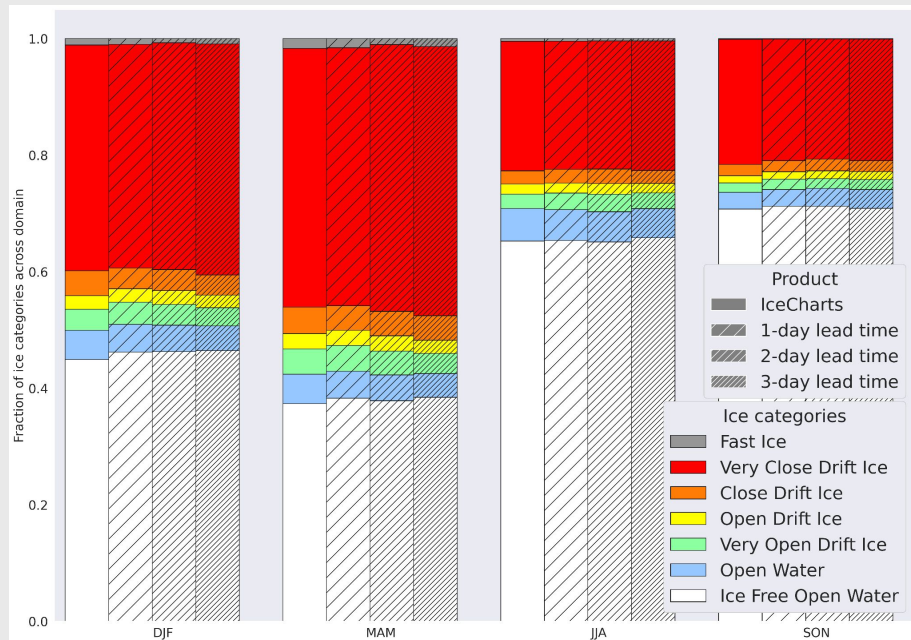
$$c_{i,j}^n = \begin{cases} 1 & \text{if } s_{i,j} \geq k_n \\ 0 & \text{if } s_{i,j} < k_n \end{cases}$$

- Separate output layer for each cumulative contour (shared decoder)

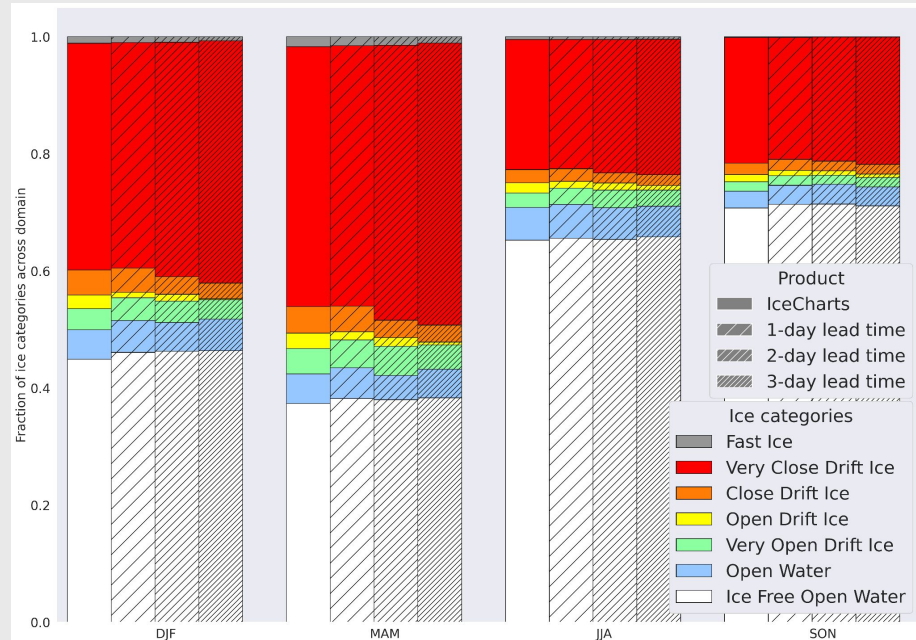
Ronneberger, O., Fischer, P., and Brox, T.: U-Net: Convolutional Networks for Biomedical Image Segmentation, in: Lecture Notes in Computer Science, pp. 234–241, Springer International Publishing, https://doi.org/10.1007/978-3-319-24574-4_28, 2015.

Target reformulation into cumulative contours

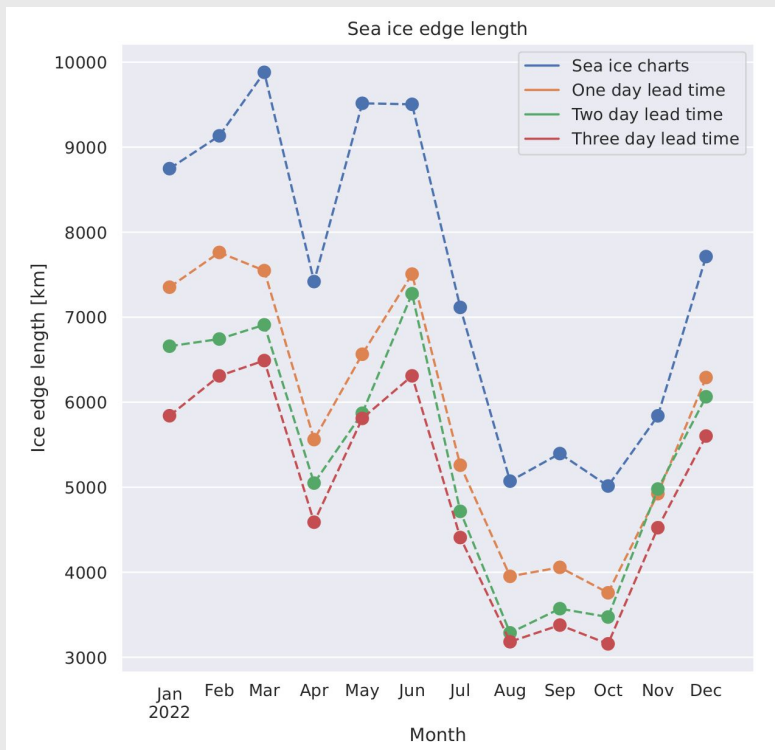
Cumulative contours



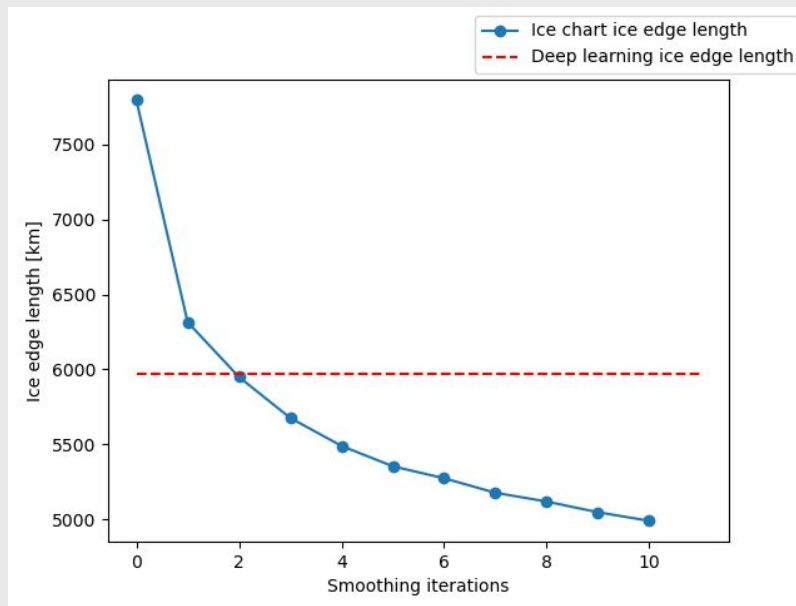
Single output, multiple labels



Predicting with the deep learning model

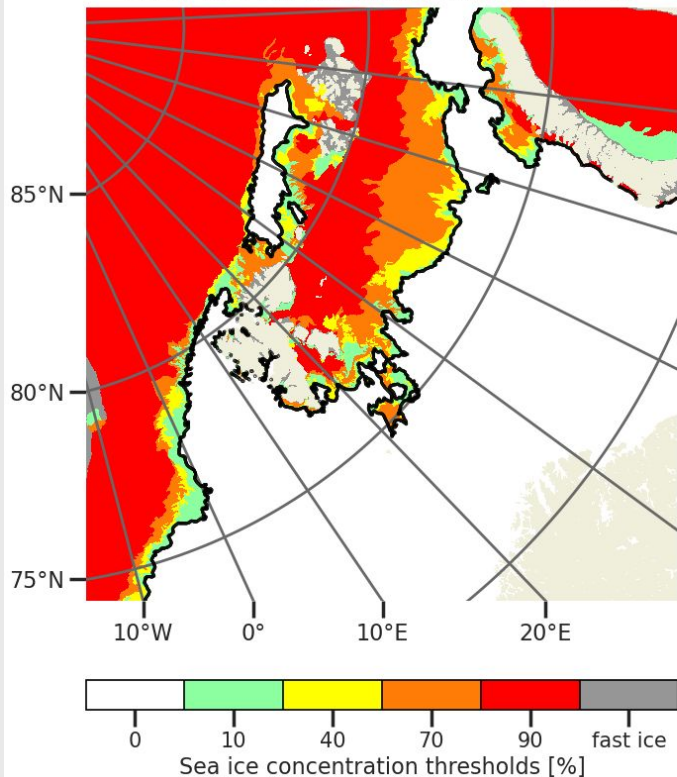


- Progressively shorter ice edge
- Blurrier forecasts at longer lead times

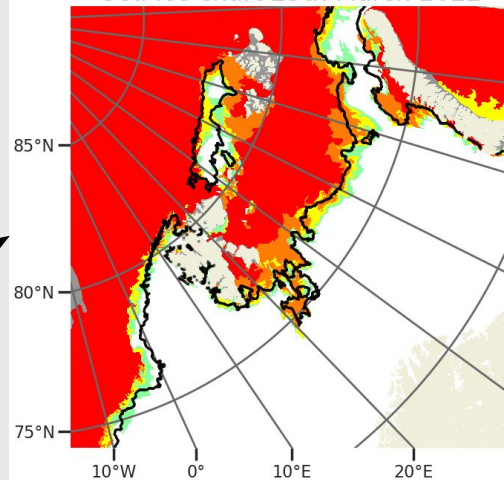


Predicting with the deep learning model

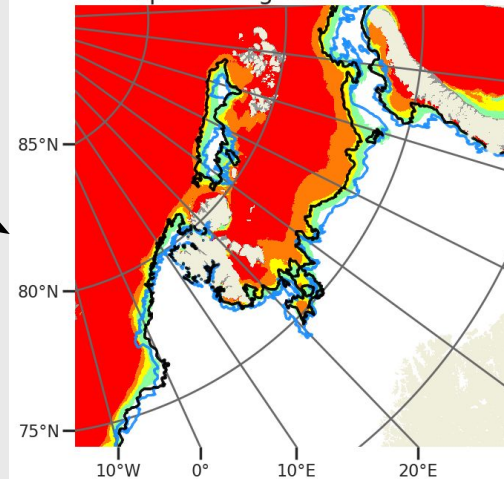
Sea ice chart 23rd March 2022



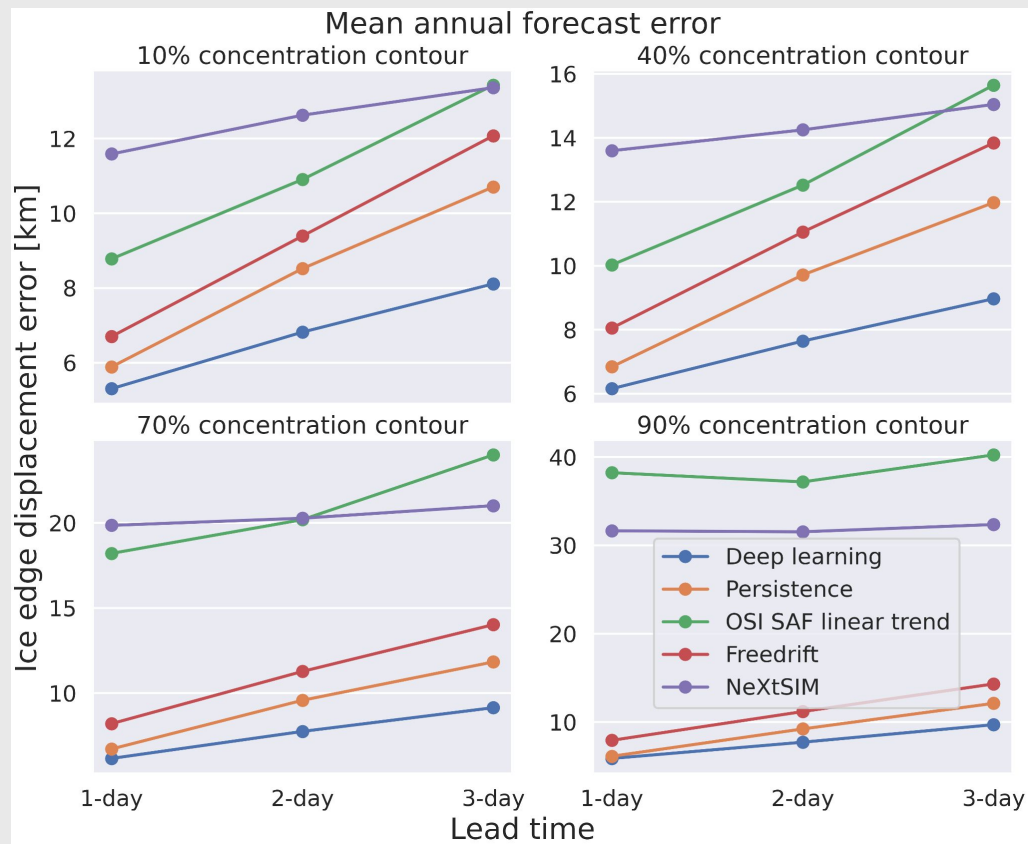
Sea ice chart 25th March 2022



Deep learning 25th March 2022



Forecast intercomparison (annual mean 2022)



- ❖ The deep learning forecasts achieves the lowest ice edge displacement error (mean) for all considered sea ice categories
- ❖ Consistent improvement against persistence, deep learning forecasts exerts some level of skill

Sub-daily regional sea ice drift forecasting with regression trees

- Predictors:
 - Buoy trajectories from IABP buoys (point based regression)
 - Wind speed -and direction from AROME Arctic forecasts
 - SIC from AMSR2 Passive Microwave observations
 - Geographic information
 - distance/angle to land
 - sector angle
- Train: 2016 - 2022, Test: 2023
- Fit Random Forest and XGBoost estimators

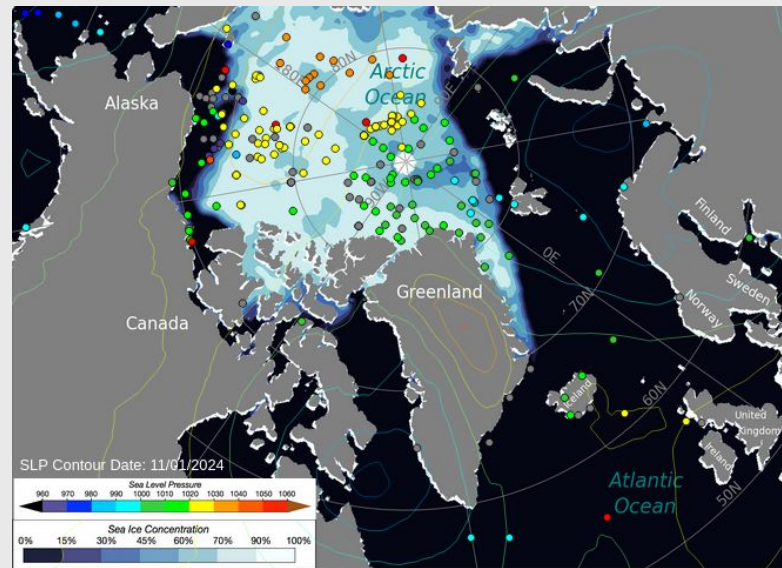
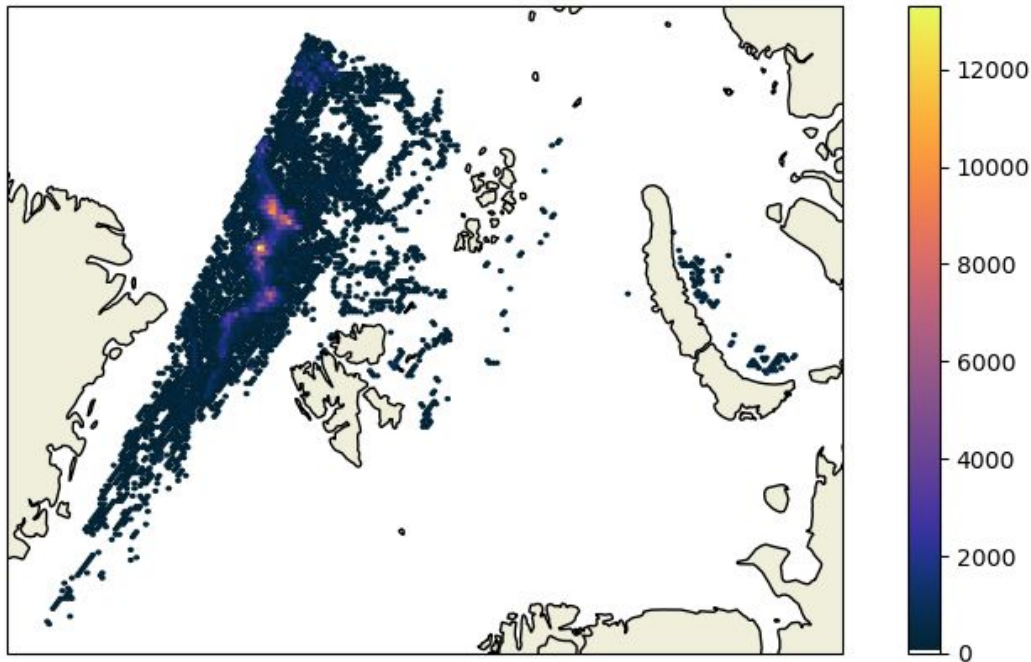


Figure from: https://iabp.apl.uw.edu/IABP_Maps.html
(Downloaded 02/11-24)

Sub-daily regional sea ice drift forecasting with regression trees

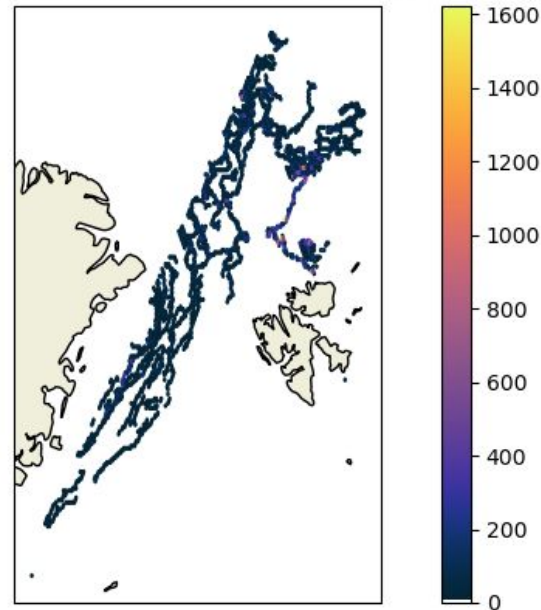
Training samples (2016 - 2022)

Total number of samples: 2203710



Test samples (2023)

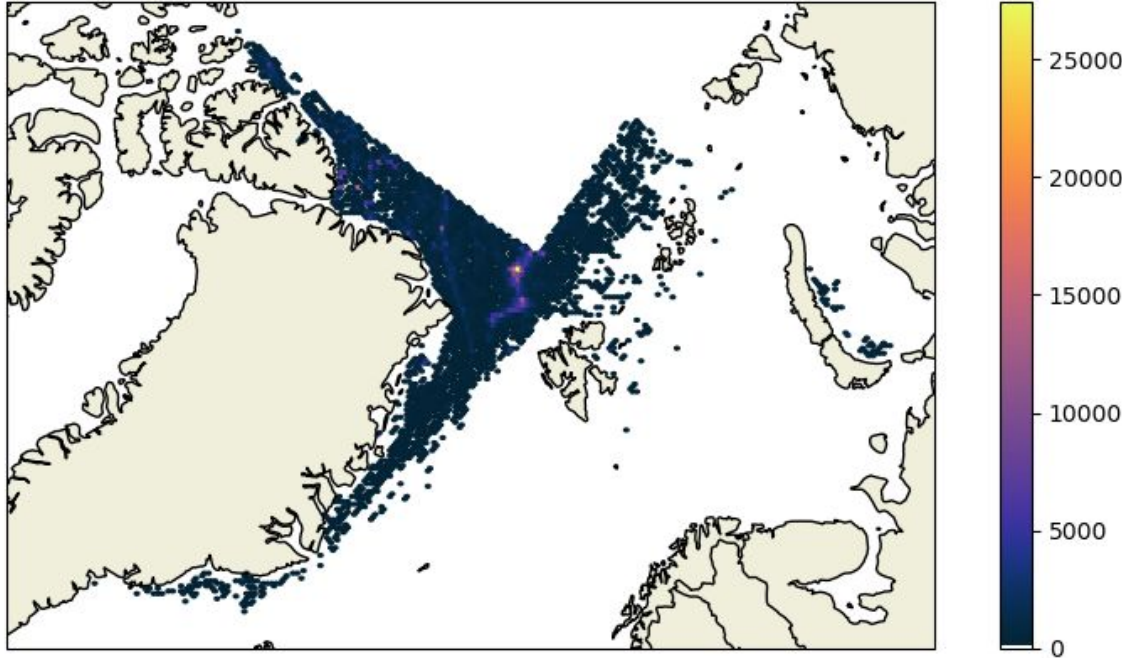
Total number of samples: 159879



Sub-daily regional sea ice drift forecasting with regression trees

Regional reanalysis

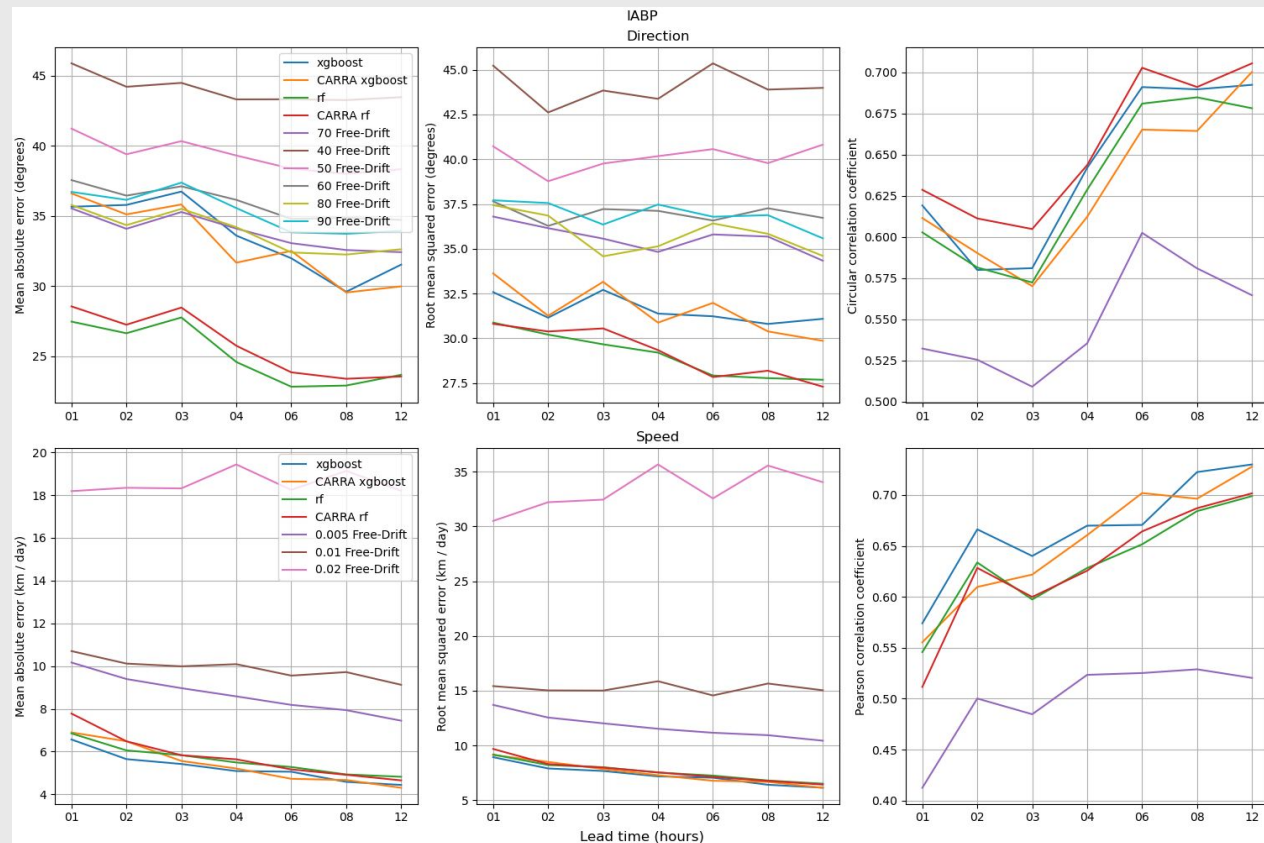
Total number of samples: 2879700



- Replace AROME Arctic and Barents domain with Copernicus Arctic Regional Reanalysis
- Extend training period (2010 - 2023) and domain (CARRA West + East)

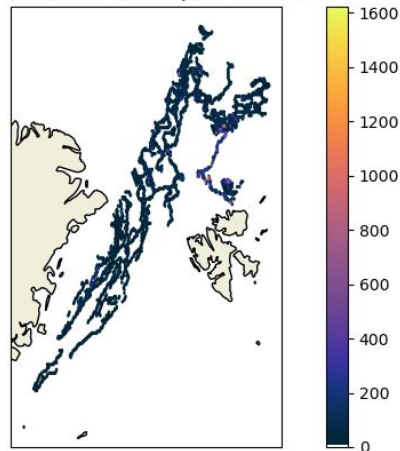
Sub-daily regional sea ice drift forecasting with regression trees

Preliminary results



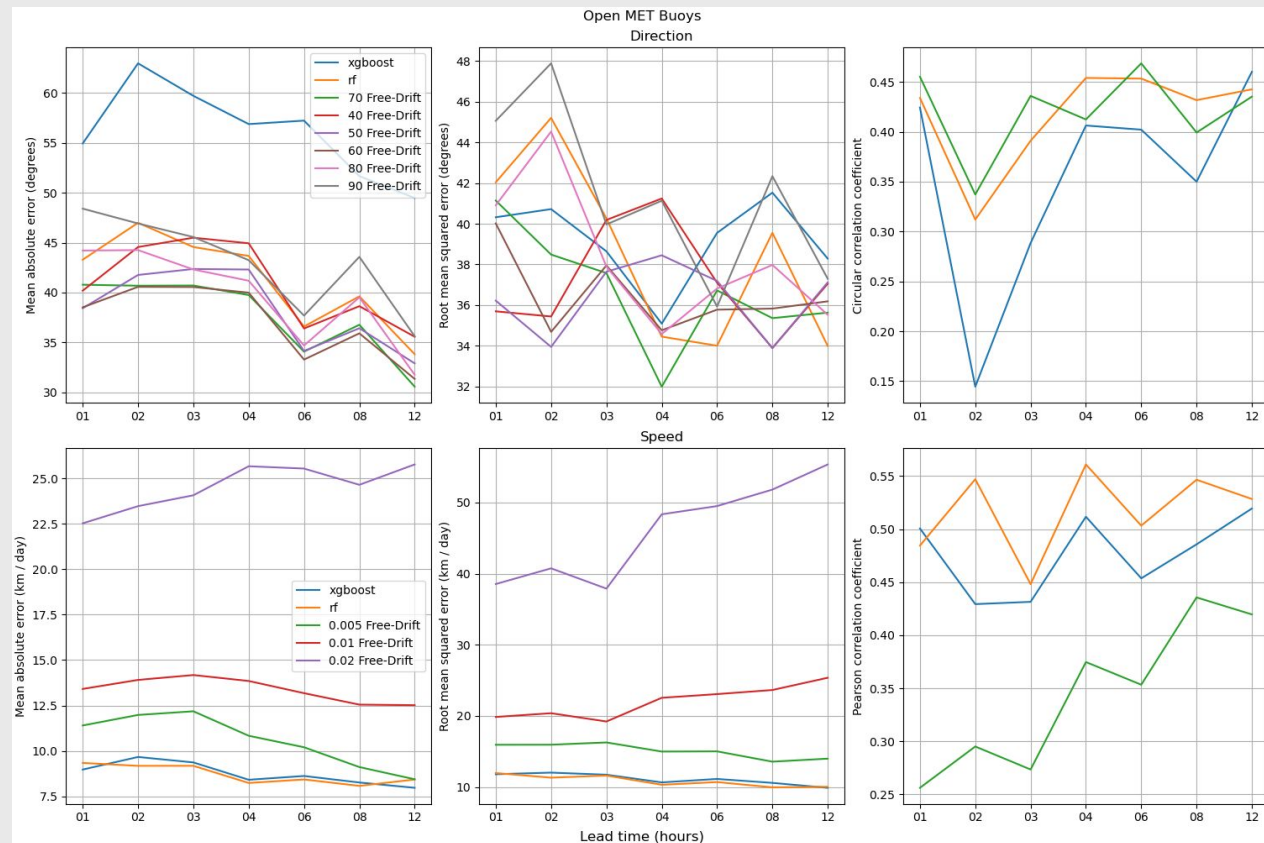
Test samples (2023)

Total number of samples: 159879

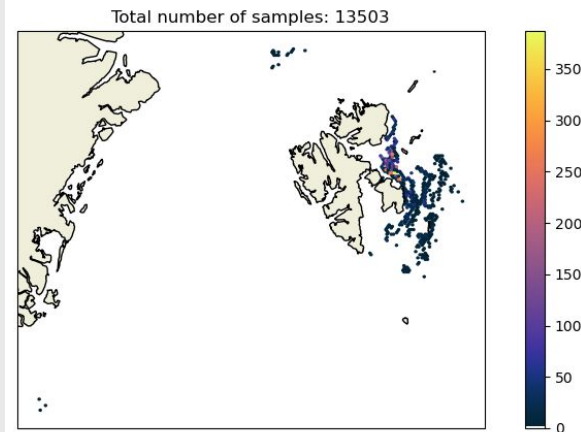


Sub-daily regional sea ice drift forecasting with regression trees

Comparison with OpenMETBuoys

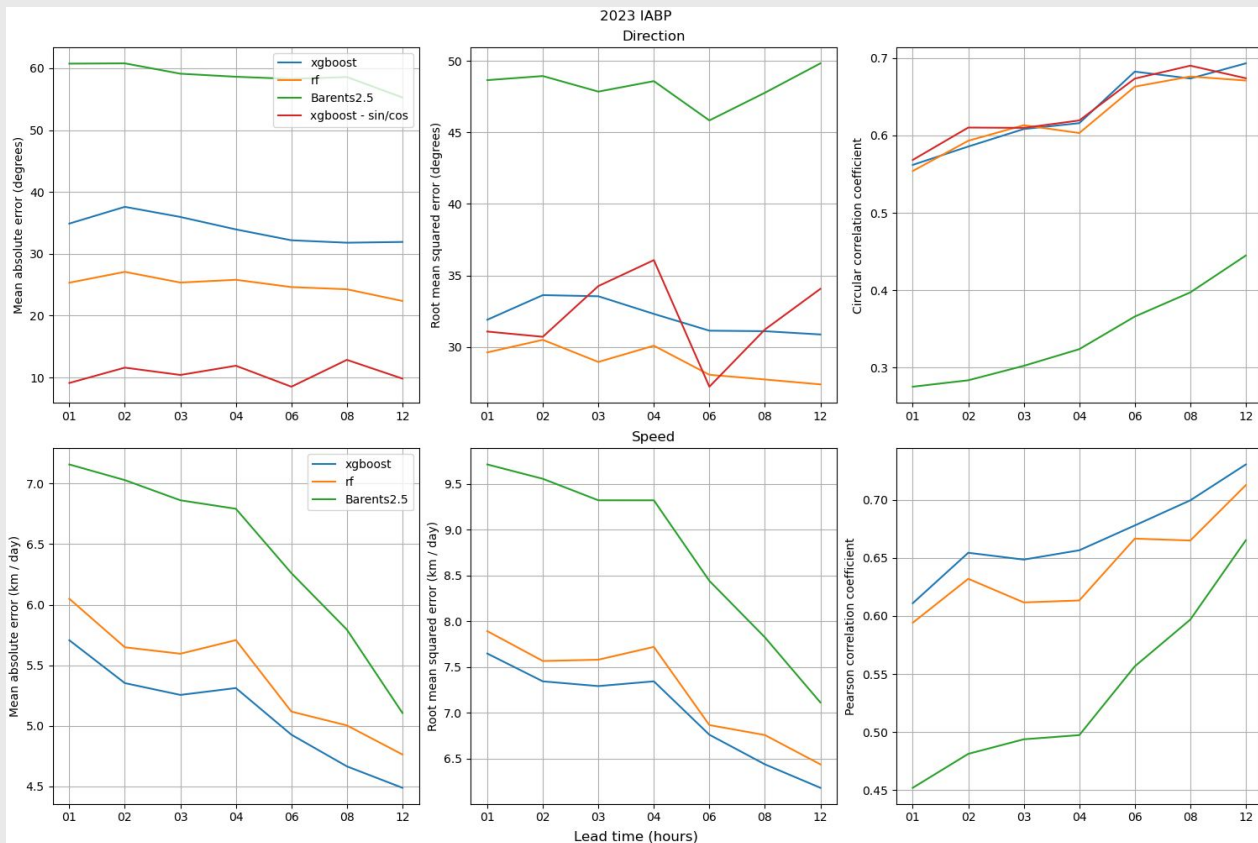


OpenMETBuoys
(2018 - 2022)



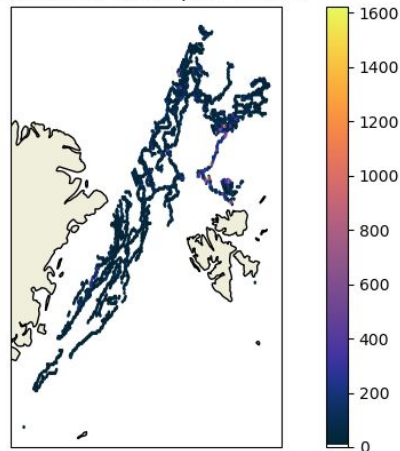
Sub-daily regional sea ice drift forecasting with regression trees

Comparison with Barents2.5



Test samples (2023)

Total number of samples: 159879



Summary

- Lightweight machine learning based models demonstrate predictive skill considering sea ice concentration and drift for short lead times
- Input and predictions are restricted to a limited model domain
- Updates to operational forecasting systems can systematically alter input data over time
- AROME Arctic currently have a static sea ice concentration, **ongoing work** to couple numerical weather forecasts with machine learning sea ice forecasts to achieve a dynamic sea ice cover

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doi: 10.5194/egusphere-2023-3107 (Kvanum 2024, under review)

