ALFRED-WEGENER-INSTITUT HELMHOLTZ-ZENTRUM FÜR POLAR-UND MEERESFORSCHUNG



# Reference forecast of sea-ice edge using damped persistence of probability anomaly

1VMW-0, 2020

**Bimo**chan Niraula, Helge Goessling Alfred Wegener Institute



## Motivation

- Operational ice-forecasting centers are producing sub-seasonal to seasonal (S2S) forecasts, whose <u>prediction skill</u> needs to be properly assessed.
- To determine the skill, the Forecast output is measured against 'truth', for example by using the <u>Spatial Probability Score</u> (SPS)

$$SPS = \int_{V} \left\{ P_{f}(\mathbf{x}) - P_{o}(\mathbf{x}) \right\}^{2} dV$$

- This score is compared against the score of some <u>reference forecasts</u>. Most common references are:
  - Climatology (historical pattern for that date) and
  - Persistence (assume initial state to stay constant).





(Zampieri et al, 2018)



# Knowledge Gap

Issues with Climatology and Persistence

- take information from either historical or current state, ignoring other.
- compare grid to grid so what happens in the neighbourhood is ignored.

So, is it possible to have a better <u>reference forecast</u> for other dynamical models to compare against?

### Yes!



### Method – 1D example



#### Notes:

i) We are using OSI SAF ice-concentration data (15% contour) to determine ice-edge.

ii) Climatology is based on the previous 10 years.



Legend:

--- Climatology (10%, 50%, 90% probability of historical ice presence)

- --- Observation (15% ice-concentration contour)
- forecast (background color = forecasted ice-probability from 0 to 100)



#### Legend:

- --- Climatology (10%, 50%, 90% probability of historical ice presence)
- --- Observation (15% ice-concentration contour)
- forecast (background color = forecasted ice-probability from 0 to 100)



#### Spatial Probability Score





#### 1.00 က b) Antarctic a) Arctic 0.75 2 SPS (10<sup>6</sup> km<sup>2</sup>) 0.50 0.25 ECMWF ECMWF\_PRES ECMWF ECMWF\_PRES UKMO DampAnomalyPerst UKMO DampAnomalyPerst KMA **DetAnomalyPerst** KMA DetAnomalyPerst NCEP Climatology NCEP Climatology CMA CMA Persistence Persistence 0.00 MFRANCE MFRANCE $\bigcirc$ 0 10 20 30 40 50 60 0 10 20 30 50 60 40 Leadtime (days) Leadtime (days)

#### Skill compared to other models in S2S dataset

Arctic Sea Ice Extent 2019

We can also look at sea-ice extent using our forecast. Here, we show the 30 day forecasts initialized at the start of each month in 2019, compared to the actual observed area.





# Other verification metrics

#### Integrated Ice Edge Error



#### Modified Haussdorff Distance

Forward distance is measured from each point in one ice-edge to the other ice-edge (or coast) and backward distance is computed from the other ice-edge. MHD is the maximum among the mean forward and backward distance, in km.

150W

 $ModHaus(A,B) = max\{mean_{a\in A} d(a, B); mean_{b\in B} d(b, A)\}$ 

#### Modified Haussdorff Distance

150W

part)

120

Ο.

Determining correct distances between two contours is not trivial, especially for non-continuous contours.

Including coastlines in the edge computation stops the distances from being measured across land.

Here, we see two ice-edge contours: --- observation (15% ice-conc contour) --- forecast (50%, 90% probability of ice presence) at 20 days leadtime.



#### Modified Hausdorff Distance



## Summary

- We have developed a method to forecast ice-edge in the Arctic and Antarctic using only initial and historical ice-edge information.
- The forecasts based on this method are skillful at sub-seasonal to seasonal timescales and are a challenging benchmark for dynamical sea-ice models to compare against.
- The difference in skill of the forecasts may vary based on the verification metric used.

# Thank you – Merci – Danke 😳

#### Questions / Comments ?

Further contact: bimochan.niraula@awi.de