Assimilation of sea ice in an Earth system model and its impacts for climate prediction

Kimmritz¹, **Counillon**^{1,2}, Wang ¹, Keenlyside^{2,1}, Bethke³

1. Nansen Environmental and Remote Sensing Center, Bjerknes Centre for Climate Research, Norway

- 2. Geophysical Institute, University of Bergen and Bjerknes Centre for Climate Research, Norway
- 3. Uni Research Climate, Bjerknes Centre for Climate Research, Bergen, Norway

Mail: <u>Madlen.kimmritz@nersc.no</u>



Summary

NorCPM is complemented with assimilation of sea ice concentration for S2S and S2D predictions
It is best to update the multicategory sea ice state and to carry strongly coupled data assimilation of ocean and sea ice

Assimilation of SIC reduces error of SIT and improves the prediction skill of sea ice extent in all regions
The largest impact is seen for regions that are semi enclosed (e.g. Kara Sea, Hudson Bay) and regions where sea ice is exported from the Arctic (e.g. Greenland Sea)
There may be light degradations in the some oceanic region but this need to be further investigated

3. Added value of sea ice for Reanalysis

We compare the performance of reanalysis based on **FREE**, **NorCPMV1** and **NorCPMV2** for the period1985-2010, for SIT we use the idepedent ICESAT data set and for heat and salt content we use the EN4 objective analysis.



V1: Assimilation of ocean observation reduces
 RMSE of SIC in regions influenced by the ocean
 V2 :Assimilation of SIC (V2) further reduces the
 RMSE everywhere

V1: Assimilation of ocean observation reduces error of SIT in region of multicategory sea ice
V2: Assimilation of SIC further reduces the RMSE for first year ice regions

V1: Assimilation of ocean observations controls well the variability of the 200 m heat and salt content (HC200 & SC200)
 V2: Assimilation of SIC may degrades the performance in the Gulf Stream and the Tropical Atlantic

NorCPM system & Experiment



4. Seasonal predictions skill

Prediction skill is tested by retrospective forecast in the period from 1985 to 2010 with 9 members and 4 start dates per year; we compare NorCPMV1 and NorCPMV2 which depicts the added value of SIC assimilation





We look at the detrended correlation of prediction of sea ice extent (SIE=sum(area | SIC > 15%)) vs that calculated from HadiSST2 for each lead-month (y-axis) and calendar month (x-axis) and organised by start date. A black dot indicates that the correlation is not significant



NorESM-O: isopycnic coordinate ocean model with bulk mixed layer on top, based on MICOM

SST+sea ice conc (SIC): HadiSST2
T-S profiles: EN4 objective analysis

Twin experiment: finding the optimal assimilation strategy to make best use of sea ice concentration

- Synthetic observations are generated from an independent pre-industrial (PI) run from NorESM
- Testing the skill of data assimilation (DA) from multiple reanalyses with monthly assim using different state vector in PI condition
- Test the optimal performance by the run that: minimise the error of the whole state (ocean, sea-ice), does not introduce a bias and preserves reliability

It is found that :

1. Updating the multicategory sea ice state (multi) outperforms update of the aggregated sea ice state (single)



2. Preserving the prior thickness per category avoid introducing a drift in thick category and does not degrade the skill



3. Joint update of ocean & sea ice (strongly coupled DA) outperforms sea ice only update (weakly coupled DA)

ice[%]

We look at the correlation of SST at 6 months and 12 months lead time





4. Performance of the optimal setting for a longer reanalysis



RMSE of the reanalysis state variables for two decades. Performance is stable when comparing the two decades. DA of SIC reduces effectively error of SIC (60%), thickness (25%), SST (30%), SSS (17%). The reliability of the ensemble is preserved.

ocean.

Kimmritz et al. 2018

Left: Averaged performance of

the two runs for sea ice and

Right: RMSE difference

and sea ice improves

regions influenced by the

between strong and weak.

Strongly coupled DA of ocean

performance of thin sea ice in

ocean variables



Reference :

- Counillon et al. "Flow-dependent assimilation of SST in isopycnal coordinates with NorCPM. Tellus A (2016)
- Wang et al. "Optimising assimilation of hydrographic profiles into isopycnal ocean models with ensemble data assimilation", OM (2017)
- Kimmritz et al. "Optimising assimilation of sea ice concentration in an ESM with a multicategory sea ice model". Tellus (2018)
- Kimmritz et al. "Added value of sea ice assimilation for seasonal prediction in the Arctic" in prep

Acknowledgments

This study was co-funded by the Center for Climate Dynamics at the Bjerknes Center, the Norwegian Research Council project SNOWGLACE (244166), SFE (270733) and INES (grant 270061), Bergen Research Foundation project BCPU and Nordforsk ARCPATH (76654). CPU and storage has been provided by UNINETT Sigma2 (nn9039k, and NS9039K)

- There is no big difference between the 2 systems for seasonal prediction of SST.
- Skill in the Nordic Seas and Barentsseas seems slightly improved, butdegraded in the subpolar gyre region.