A satellite image of the Arctic region showing sea ice patterns. The ice is depicted in various shades of white and light blue, with darker blue areas representing open water. The ice is broken up into numerous floes and leads, creating a complex, fractal-like pattern. The overall appearance is that of a vast, textured expanse of frozen water.

Prediction of Arctic sea ice on subseasonal to seasonal time scales

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ECMWF Seminar

September 15th 2017

- Research Motivation and Objectives
- S2S Forecasts and Observations
- The Verification Metrics
- Predictive Skills of S2S Forecasts Systems
- Comparison of Predictive and Prescriptive Systems
- Considerations on Metrics Behavior

Research Motivations and Objectives

Why do we need (Arctic) sea ice forecasts?

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Climate change causes a decrease in summer sea ice **extent** and **thickness**



New scenarios for human activities in the Arctic region

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New scenarios for human activities in the Arctic region

- Marine transport
- Offshore fuel industry
- Mineral extraction
- Tourism

Why do we need (Arctic) sea ice forecasts?

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New scenarios for human activities in the Arctic region

- Marine transport
- Offshore fuel industry
- Mineral extraction
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Formulation of seasonal sea ice forecasts is required

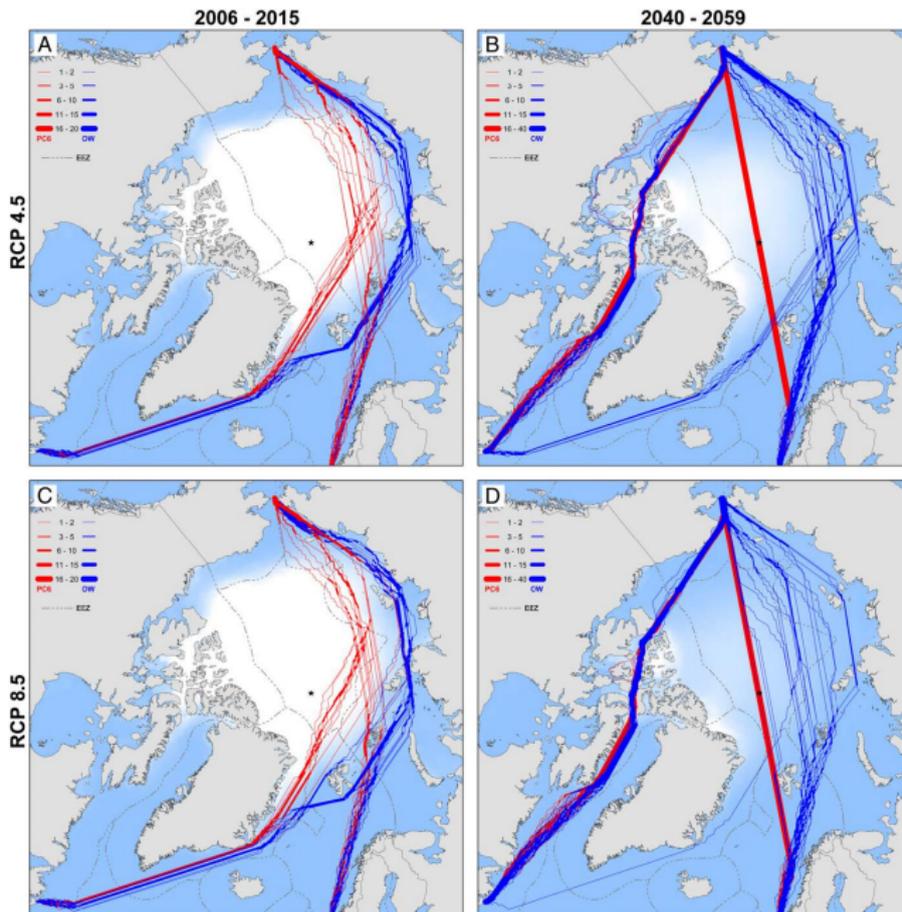


Figure: Hypothetical September navigation routes. Smith and Stephenson (2013)

Are we able to effectively verify a sea ice forecast?

Are we able to effectively verify a sea ice forecast?



New dedicated **verification metrics** are needed to quantify the **quality** of the forecasted **ice edge position**

This research consists in an extensive verification analysis of the S2S database with the following objectives:

- **Assessment of the predictive skills for S2S forecast systems**
- **Evaluation of the verification metrics behavior**

Forecasts and Observations

The S2S (subseasonal to seasonal) database collects mainly **atmospheric forecasts** (2003-2017). However, sea ice concentration is also provided.

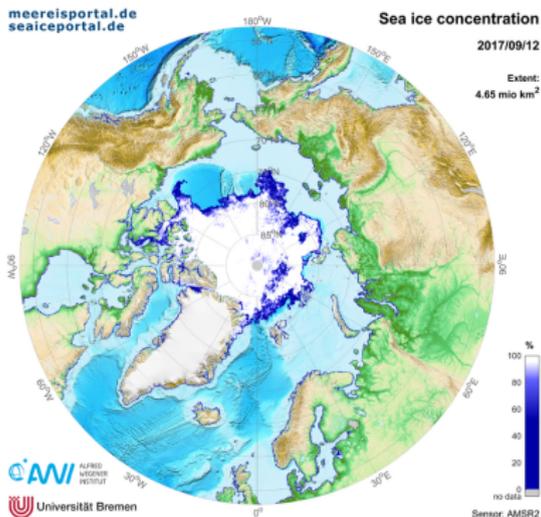
F. Vitart et al. (2017)

The S2S (subseasonal to seasonal) database collects mainly **atmospheric forecasts** (2003-2017). However, sea ice concentration is also provided.

Model Name	Ocean	Sea Ice	Frequency	Ens. Size	Length
BoM	✓		twice a week	33	62 days
ECCC			weekly	21	32 days
ECMWF 1			twice a week	51	46 days
HMCR			weekly	20	61 days
ISAC-CNR			weekly	41	31 days
JMA			twice a week	25	33 days
CMA	✓	✓	daily	4	60 days
ECMWF 2	✓	✓	twice a week	51	46 days
KMA	✓	✓	daily	4	60 days
Météo France	✓	✓	weekly	51	32-61 days
NCEP	✓	✓	daily	16	44 days
UKMO	✓	✓	daily	4	60 days

ASI sea ice concentration data
produced by **University of
Bremen**.

The resolution is ~ 6 km.



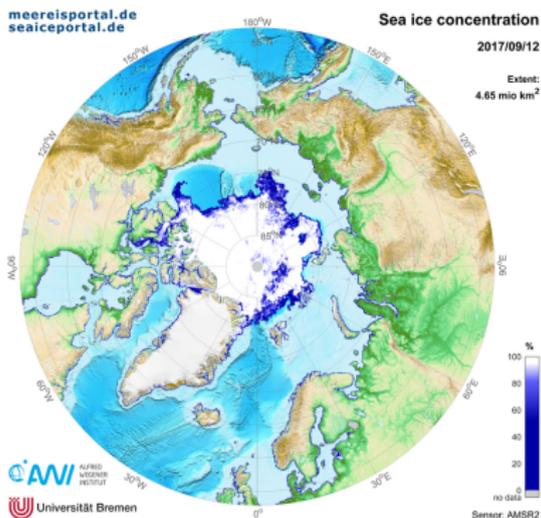
G. Spreen et al. (2008)

ASI sea ice concentration data
produced by **University of
Bremen**.

The resolution is ~ 6 km.

Models own analysis

The idea behind the models own
analysis is to define virtual
observations based on the control
forecasts evaluated at the initial
time of each single forecast.

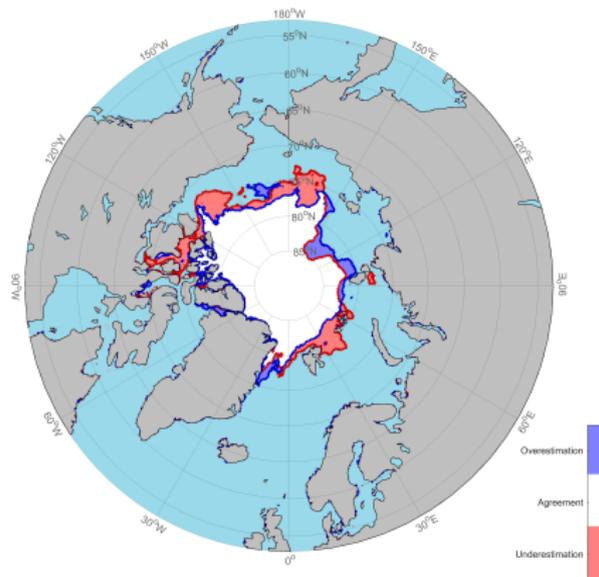


G. Spreen et al. (2008)

Verification Metrics

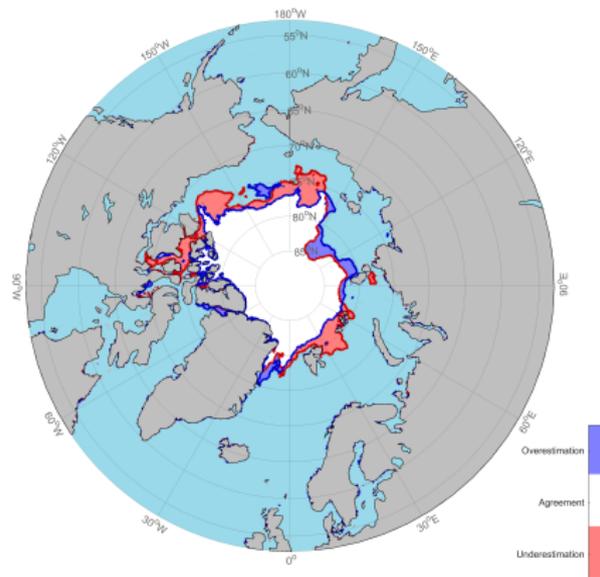
H.F. Goessling et al (2016)

- Observation edge
- Forecast edge



$$IIEE = O + U$$

- Observation edge
- Forecast edge



$$IIEE = O + U$$

- Conceptually simple and easy to calculate from sea ice concentration

- IIEE is an area (m^2)

- Decomposition into **Misplacement Error**

$$ME = 2\min(O, U)$$

and

(Absolute) Extent Error

$$AEE = |O - U|$$

$$EE = O - U$$

$$IIEE = AEE + ME$$

SPS is the evolution of **IIEE** in the probabilistic forecasts world.
SPS is defined as the spatial integration of the local (Half) Brier Score.

$$SPS = \int_S (p_o [sic \geq 15\%](\vec{x}) - p_f [sic \geq 15\%](\vec{x}))^2 dS$$

- SPS can be applied to deterministic forecast, in this case $SPS = IIEE$
- It allows a probabilistic description of the observations
- SPS is an area (m^2)
- Dividing the SPS (or the IIEE) by the climatological length of the edge we obtain an estimation of the mean distance between the edges

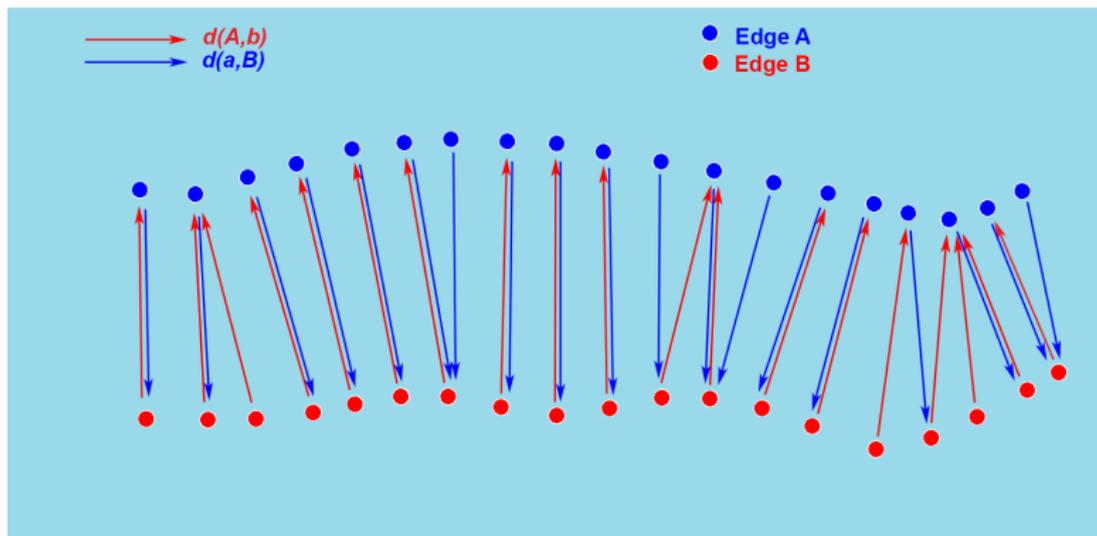
H.F. Goessling (submitted)

$$\text{MHD}(A, B) = \max \left\{ \frac{1}{|A|} \sum_{a \in A} d(a, B), \frac{1}{|B|} \sum_{b \in B} d(A, b) \right\}$$

$$d(a, B) = \inf_{b \in B} [d(a, b)]$$

$$d(A, b) = \inf_{a \in A} [d(a, b)]$$

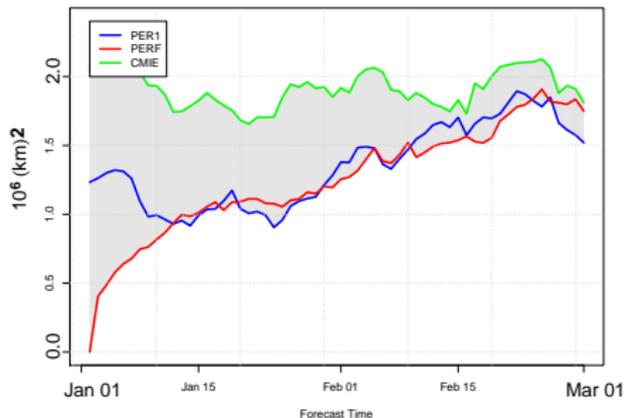
D.S. Dukhovskoy et al. (2015)



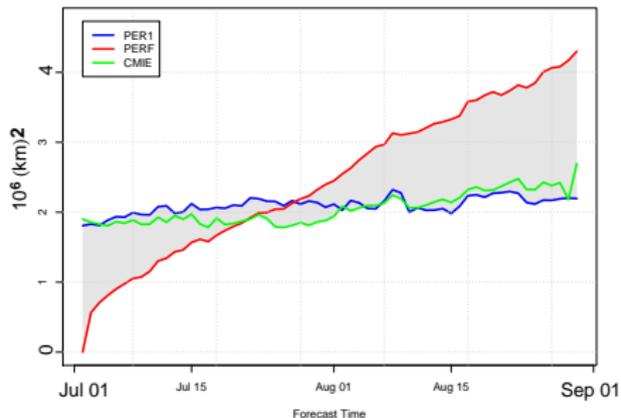
IIEE and SPS are not straightforward to interpret without reference values. Those have been calculated using the observed sea ice concentration

- **Persistence from the previous year** (PER1)
- **Persistence from forecast beginning** (PERF)
- **Climatological median ice edge** (CMIE)

Benchmark values for IIEE and SPS from AMSR2 data



Benchmark values for IIEE and SPS from AMSR2 data



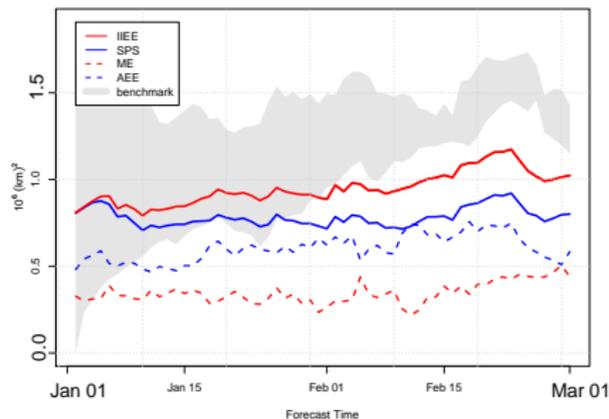
Predictive Skills of S2S Forecast Systems

Ens. members: 50
Start: 01.01.2016

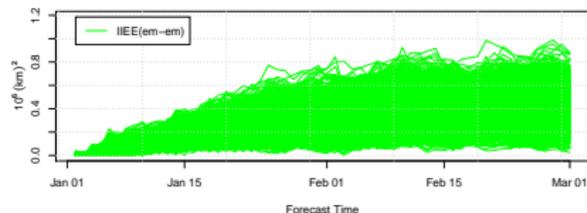
Météo France

Forc. length: 60 days

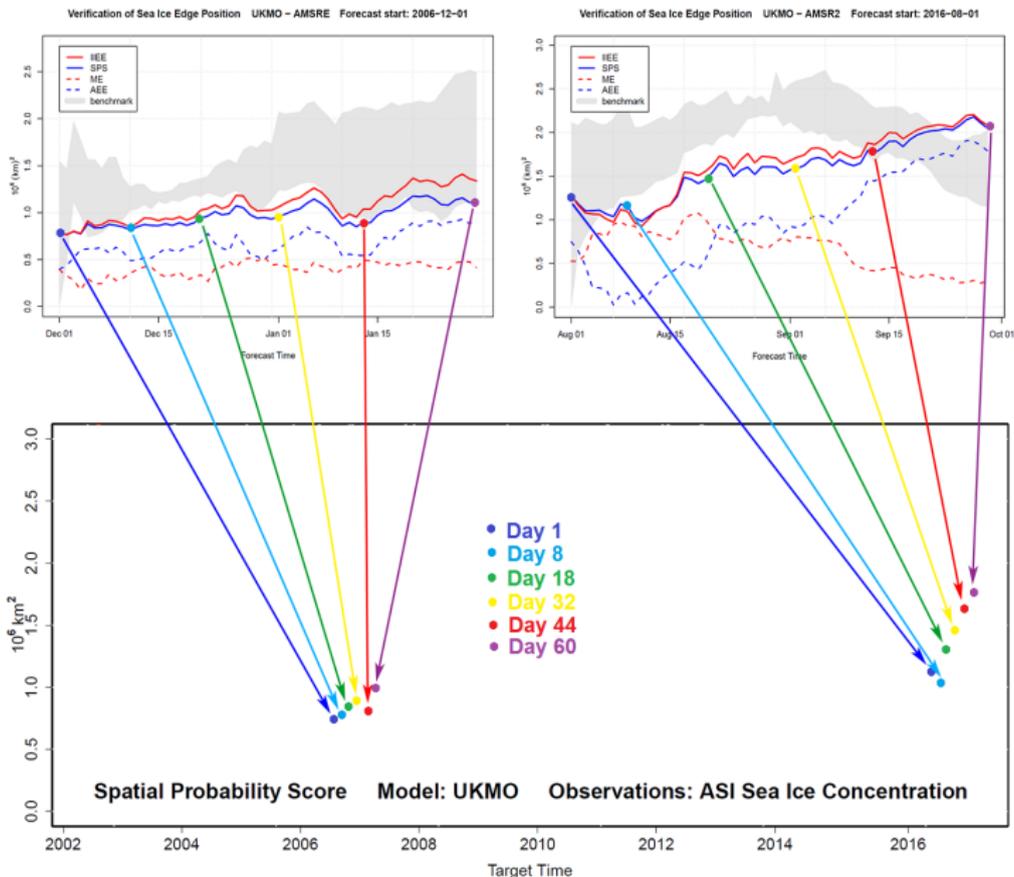
Verification of Sea Ice Edge Position Météo France – AMSR2 Forecast start: 2016-01-01

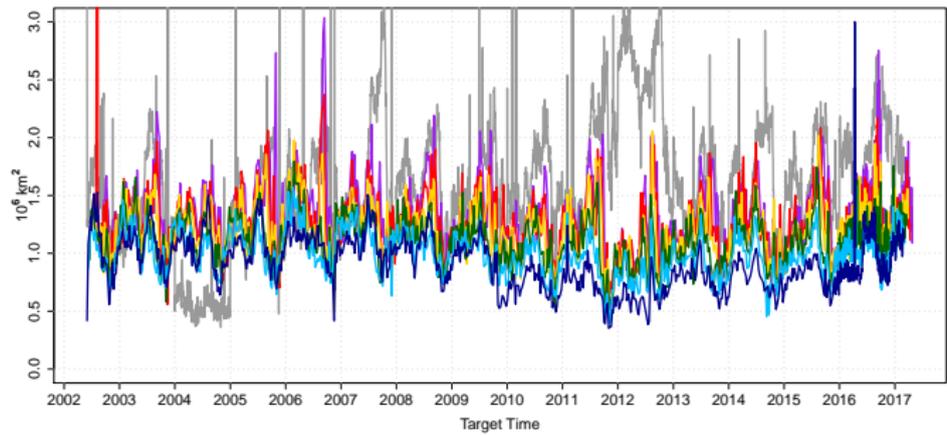


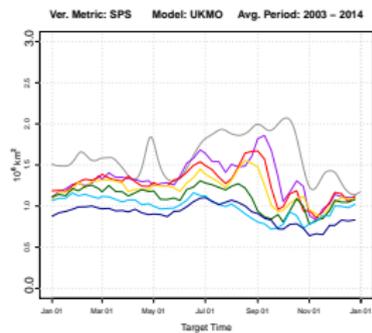
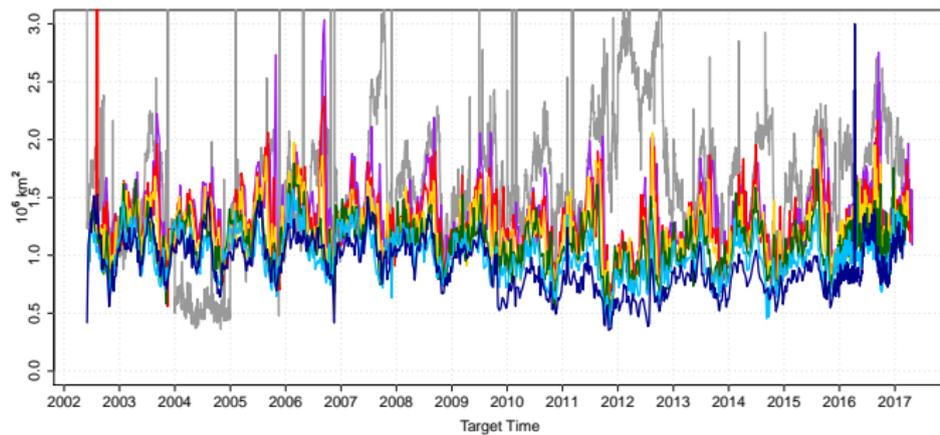
Ensemble Members Spread Météo France Forecast start: 2016-01-01

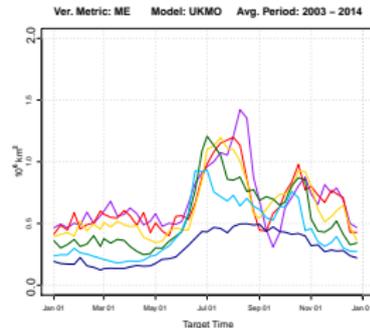
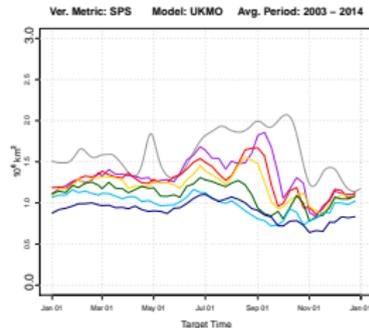
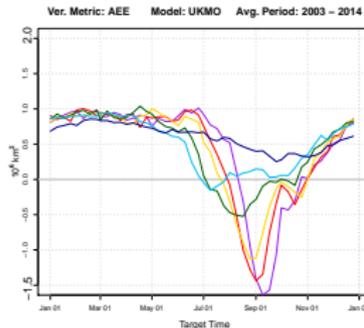
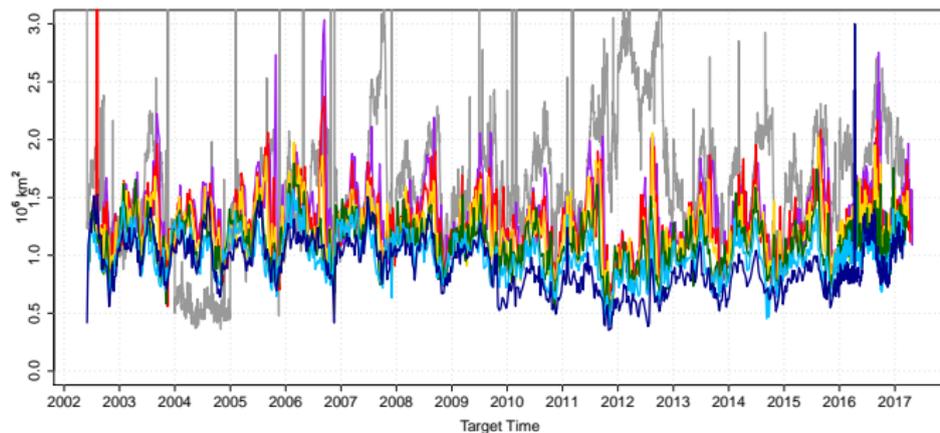


Extensive visualization of the results







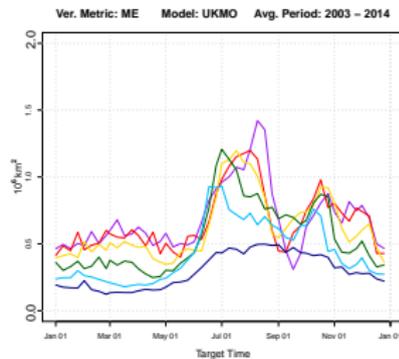
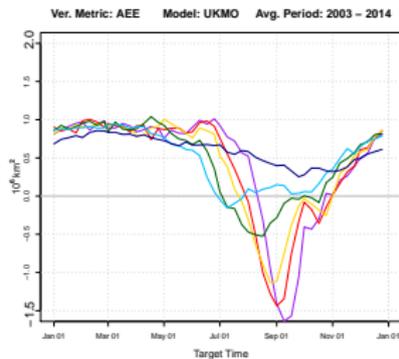
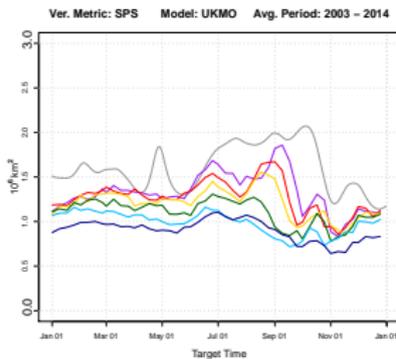


SPS

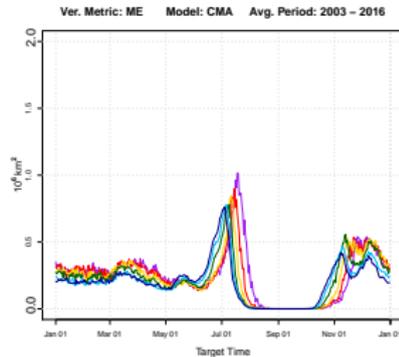
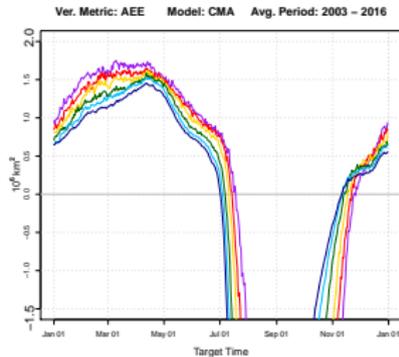
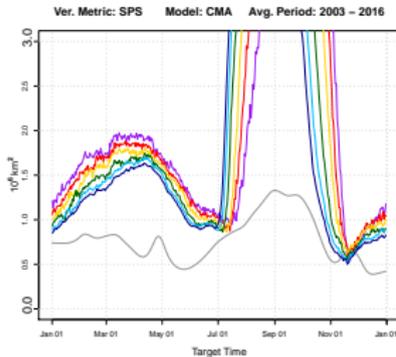
AEE

ME

UKMO



CMA



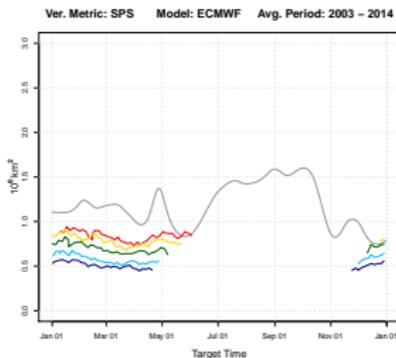
Ens. members: 3
Start: 01.07.2016 **UKMO**

Ens. members: 3
Start: 01.07.2016 **CMA**

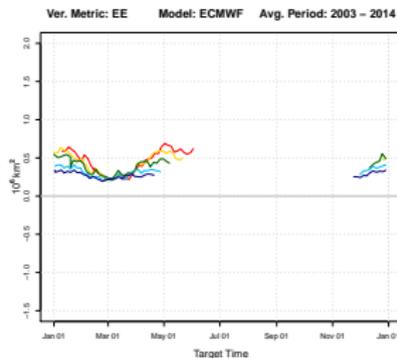
Forecast System	Season		Issues		
	Winter	Summer	Assimil.	O-Mlt.	O-Frz.
CMA	●	●	X	X	X
ECMWF 2	●				
KMA	●			X	
Météo France	●	●	X	X	
NCEP	●	●	X		X
UKMO	●	●		X	

ECMWF 2

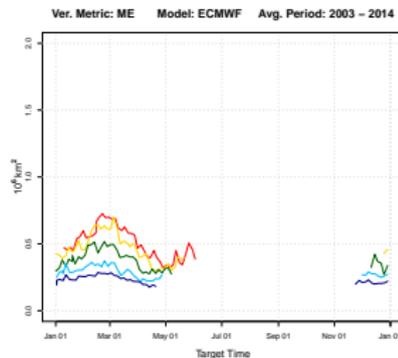
SPS



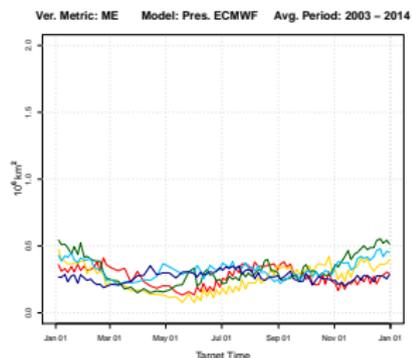
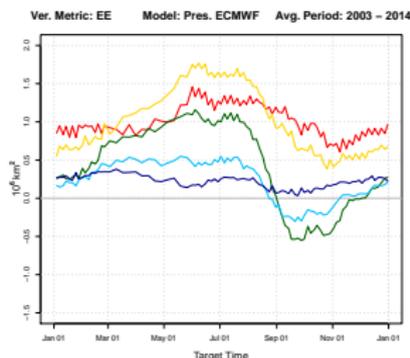
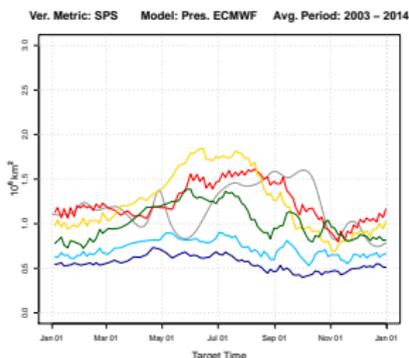
AEE



ME



ECMWF 1



Predictive Version
Start: 01.08.2016

ECMWF 2

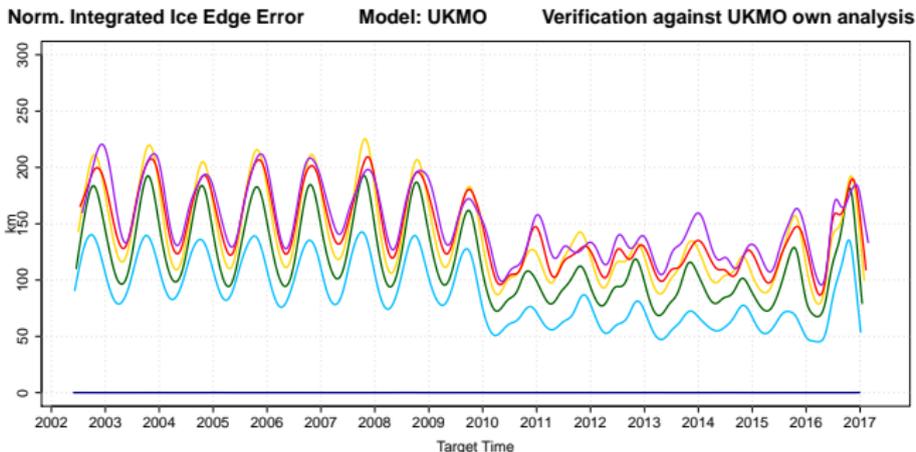
Prescriptive Version
Start: 31.07.2016

ECMWF 1

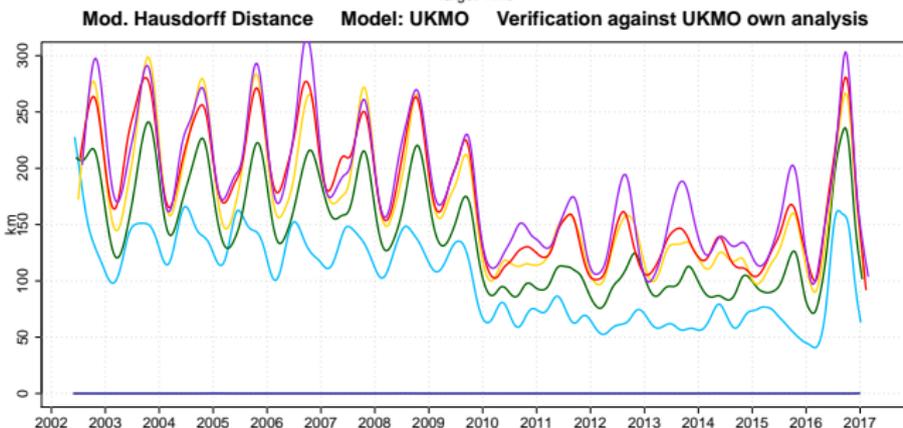
Verification Metrics Behavior

Comparison of MHD and NIIEE

NIIEE



MHD



Forecast Lead Time	Correlation Coeff.	Scaling Factor
Day 1	0.915	0.75
Day 8	0.813	1.18
Day 18	0.872	1.23
Day 32	0.860	1.24
Day 44	0.770	1.24
Day 60	0.672	1.23

The NIIEE and the MHD estimations of the mean distance between the edges are comparable! **However...**

- NIIEE is sensitive to the normalization procedure
- MHD is subject to noise likely caused by outliers
- MHD computation is much more demanding

Conclusions

- Despite the early development stage of Arctic sea ice predictions on the seasonal time scale some of the S2S models are promising, exhibiting better predictive skills than the observation-based climatology and persistence.
- Critical aspects concerning the data assimilation procedure and the tuning of the models, which can strongly affect the forecasts quality.
- Expected benefits from an increased ensemble size could not be detected.
- The comparison of different versions of the ECMWF forecast system shows the benefits brought by a coupled dynamical description of the sea ice instead of its prescription based on persistence and climatological records.

- IIEE and SPS are effective verification metrics to describe the quality of the sea ice edge position.

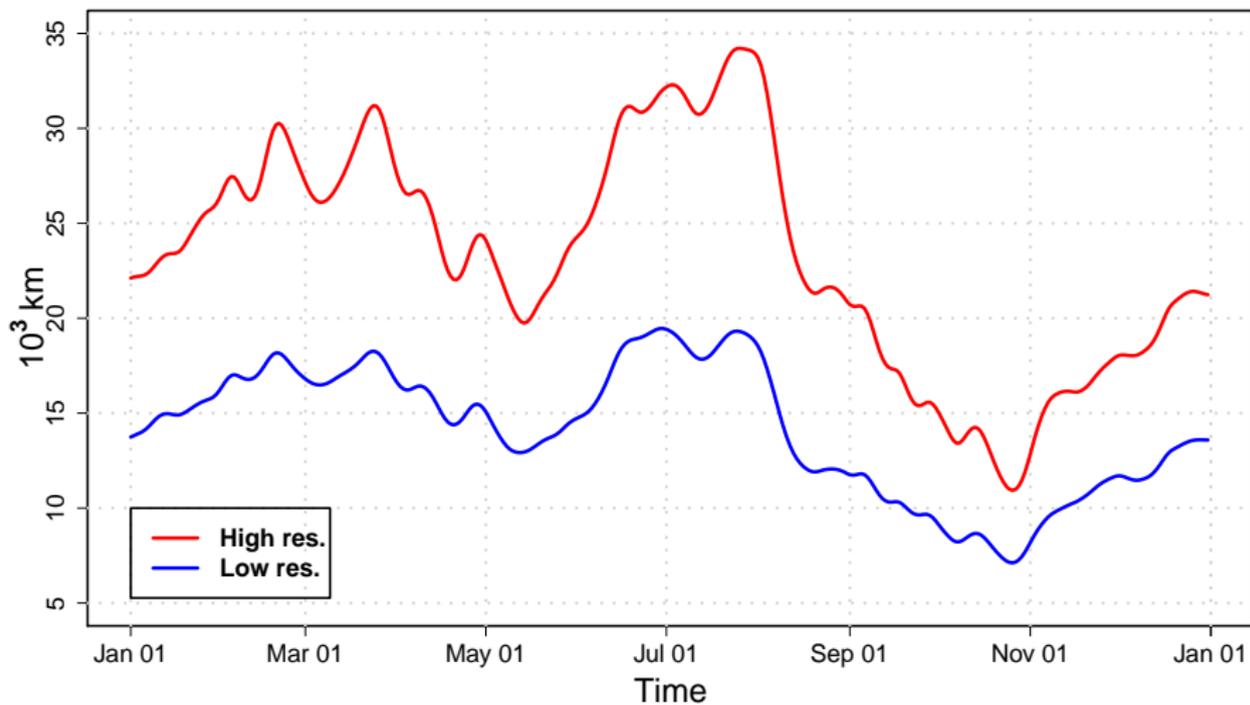
Simplicity - Comprehensibility - Stability

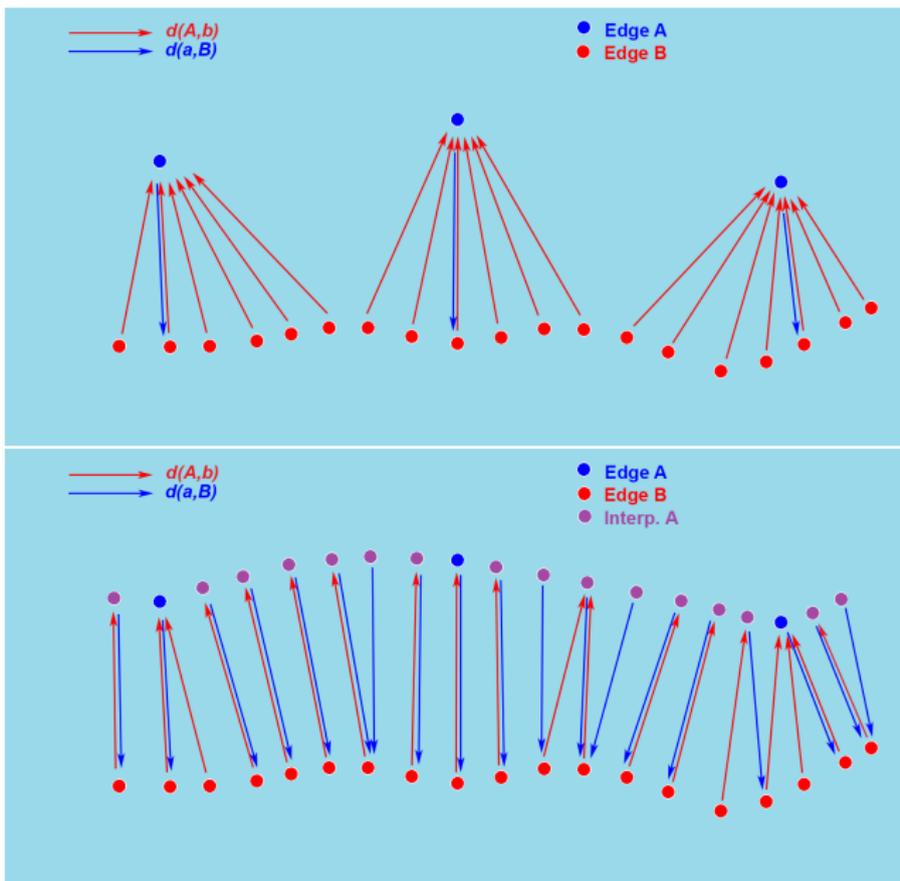
- MHD is also able to evaluate the quality of the forecasted ice edge position. However it is less flexible than the two previous ones and affected by biases.
- Verification against satellite observation useful to monitor models skills.
- Verification against models own analysis useful to study the model response to modification in data assimilation.

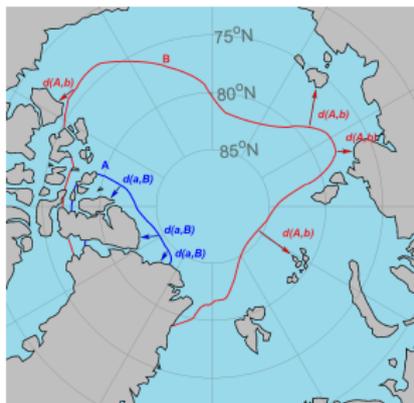
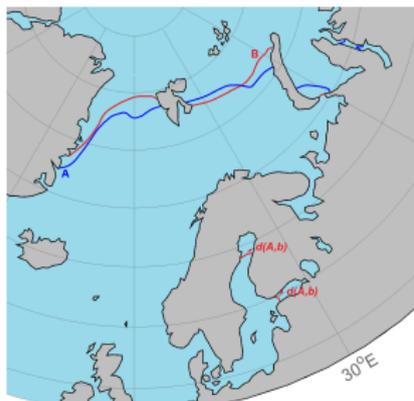
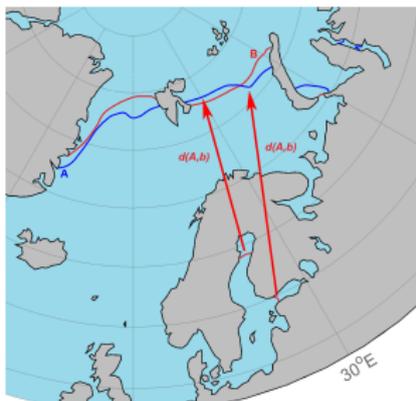


Thank you for your attention

Climatological Ice Edge Length – ASI Sea Ice Concentration







meereisportal.de

seaiceportal.de

Sea Ice Edge Comparison

2017/03 - 2006/03

