

# An Overview of European Union- Funded Project **APPLICATE**



Pablo Ortega, on behalf of APPLICATE partners

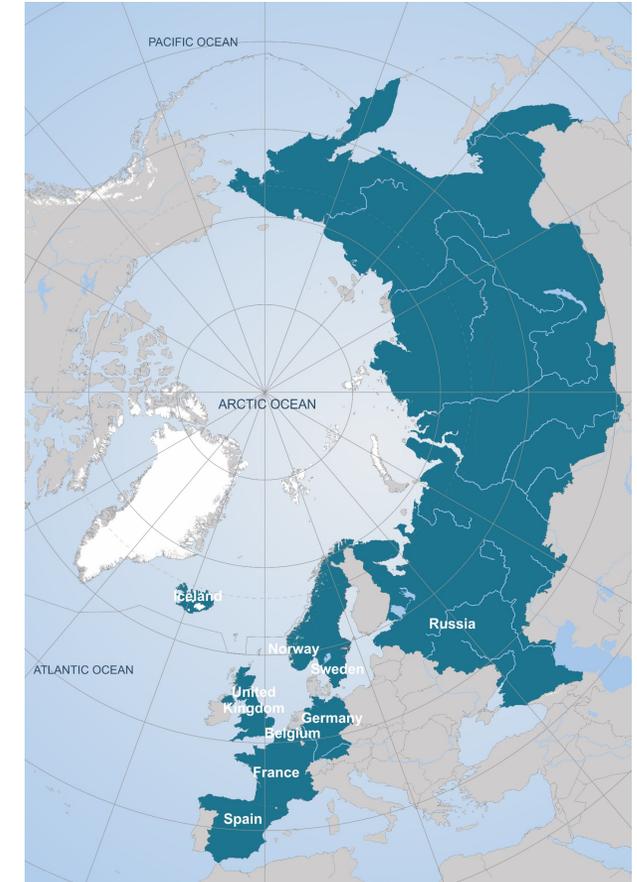
**APPLICATE.eu**   
Advanced prediction in  
polar regions and beyond



This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 727862.

# The Consortium

16 partners and 1 third-party from 9 countries



... and many collaborators!

# Budget and duration

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- € 8 Mio + separate Russian contribution
- 1<sup>st</sup> November 2016–31<sup>st</sup> October 2020 (4-years)
- 6 month no-cost extension requested



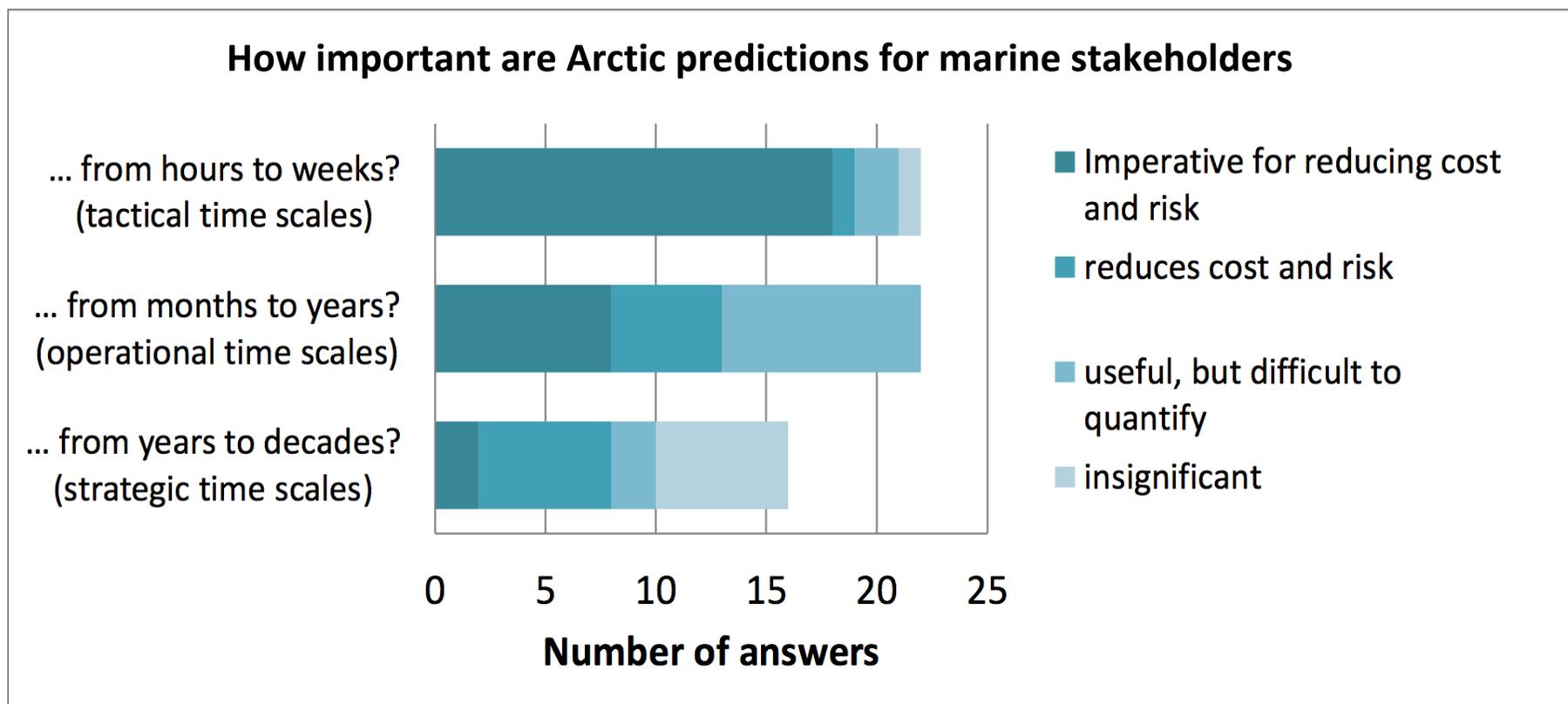
# Mission statement

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*Develop enhanced predictive capacity for weather and climate in the Arctic and beyond, and determine the influence of Arctic climate change on Northern Hemisphere mid-latitudes, for the benefit of policy makers, businesses and society.*



## ➤ Bringing together the NWP and climate communities

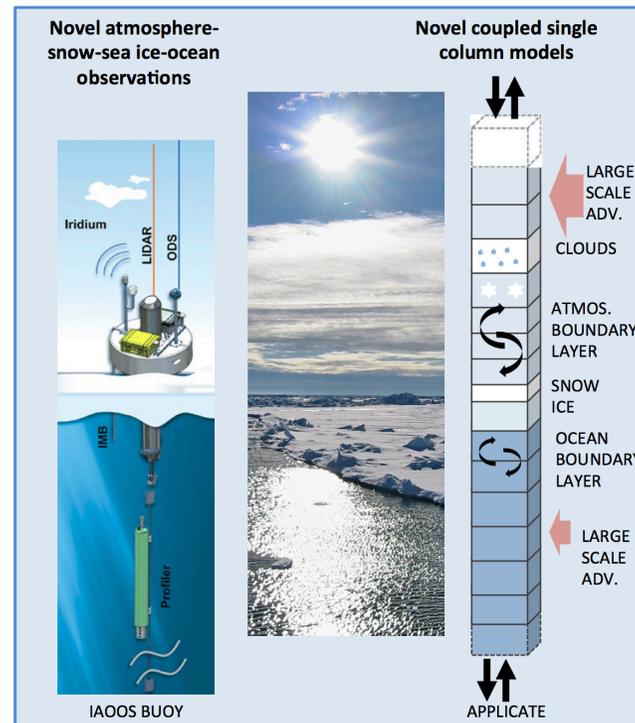


Survey: EC-PHORS Services Task Team



# General approach

- Involving experts on the Arctic and midlatitudes
- Engaging operational centres for maximizing impact
- Effectively combining models and observations



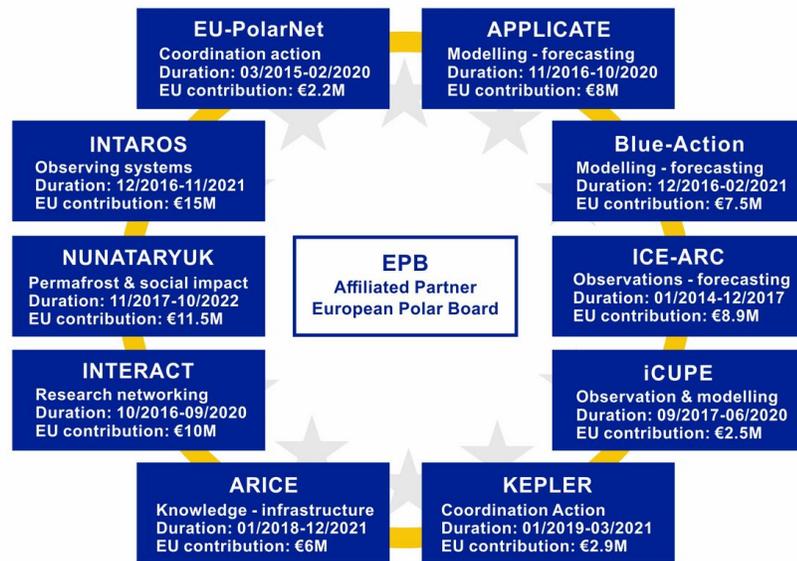
**MOSAIC**  
International  
Arctic Drift  
Expedition



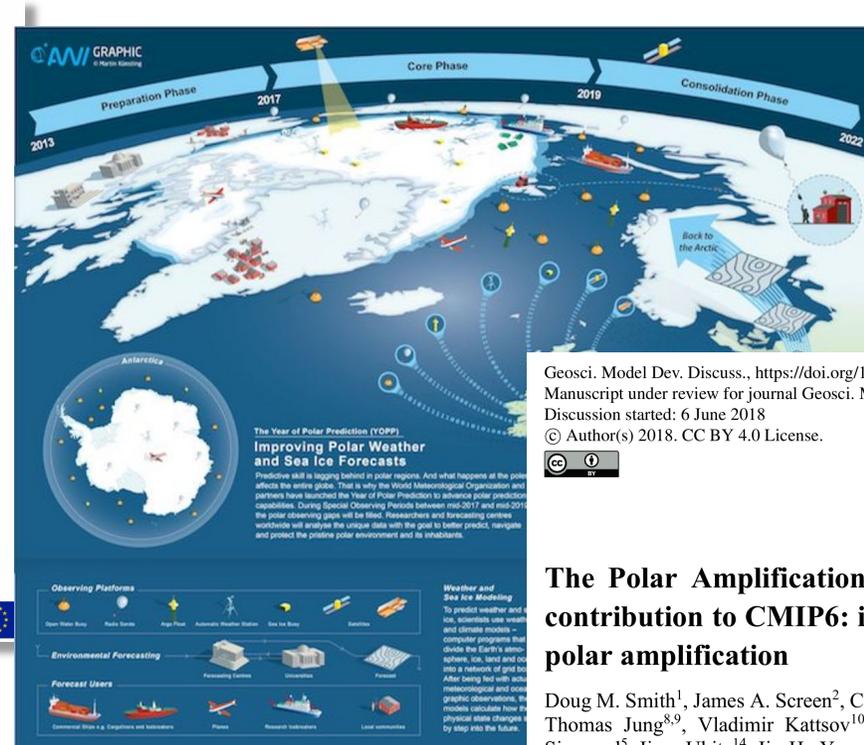
# General approach

- Exploiting European and international collaboration (e.g. Arctic Cluster, YOPP and PAMIP)

## EU ARCTIC PROJECT CLUSTER



The EU Arctic Cluster projects have received funding from the European Union's Horizon 2020 research and innovation programme or the European Union's Framework 7 Programme respectively. Please visit our website for the specific grant numbers.



Geosci. Model Dev. Discuss., <https://doi.org/10.5194/gmd-2018-82>  
Manuscript under review for journal Geosci. Model Dev.  
Discussion started: 6 June 2018  
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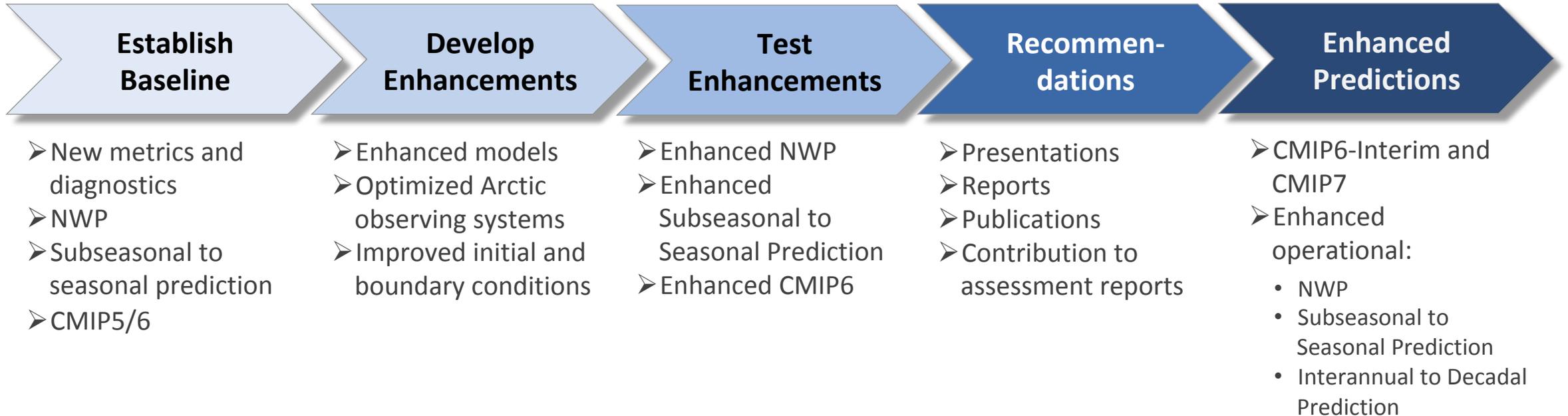
Geoscientific  
Model Development  
Discussions  
EGU

## The Polar Amplification Model Intercomparison Project (PAMIP) contribution to CMIP6: investigating the causes and consequences of polar amplification

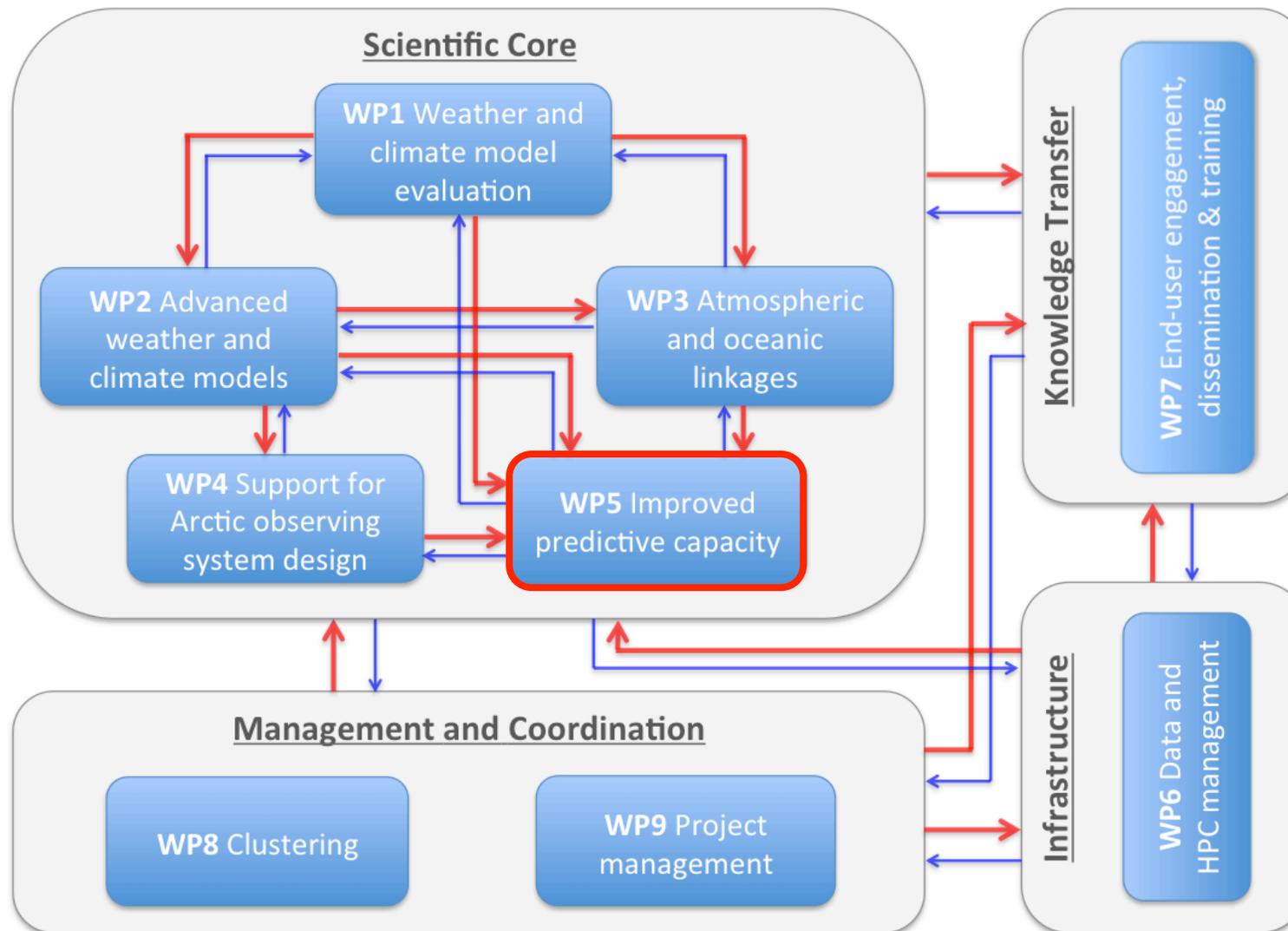
Doug M. Smith<sup>1</sup>, James A. Screen<sup>2</sup>, Clara Deser<sup>3</sup>, Judah Cohen<sup>4</sup>, John C. Fyfe<sup>5</sup>, Javier García-Serrano<sup>6,7</sup>, Thomas Jung<sup>8,9</sup>, Vladimir Kattsov<sup>10</sup>, Daniela Matej<sup>11</sup>, Rym Msadek<sup>12</sup>, Yannick Peings<sup>13</sup>, Michael Sigmond<sup>5</sup>, Jinro Ukita<sup>14</sup>, Jin-Ho Yoon<sup>15</sup>, Xiangdong Zhang<sup>16</sup>



## Delivering enhanced predictions



# Project structure



1. Advance our understanding of **predictability mechanisms** operating at 3 different timescales:

## NWP

*CRNS (CNRM)*  
*Met Norway*  
*ECMWF*

deterministic/ensemble  
global/limited area models  
focus on YOPP period

## – Seasonal Prediction

*CRNS (CNRM)*  
*Met Office*  
*BSC*  
*UCLouvain*

10 members  
1993-2014 period  
May/November ICs

## – Climate Projections

*AWI*  
*BSC*

**HiResMIP:**  
1950 fixed forcing control  
1950-2050 transient



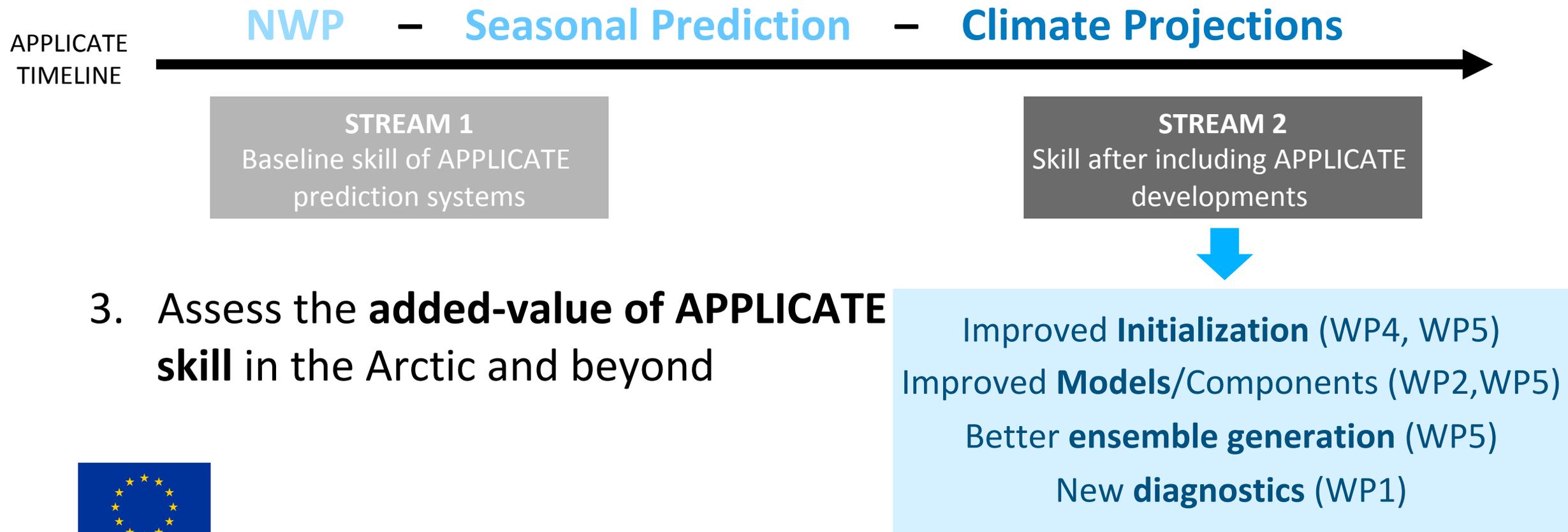
1. Advance our understanding of **predictability mechanisms** operating at 3 different timescales:

**NWP** – **Seasonal Prediction** – **Climate Projections**

2. Investigate whether and how **linkages between the Arctic and mid-latitudes contribute to prediction skill**
3. Assess the **added-value of APPLICATE developments on prediction skill** in the Arctic and beyond



# WP5: Main goals



# WP5: Main tasks

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**Task 5.1:** Production of Stream 1 experiments ✓

**Task 5.2:** State-of-the-art of weather/climate prediction and projections ✓  
(sources of predictability, links with mid-latitudes, forecasts of extremes,...)

**Task 5.3:** Added-value of improved process representation on predictive skill ✓  
(enhanced sea ice models, increased resolution, improved ensemble generation)

**Task 5.4:** Production and evaluation of Stream 2 Experiments

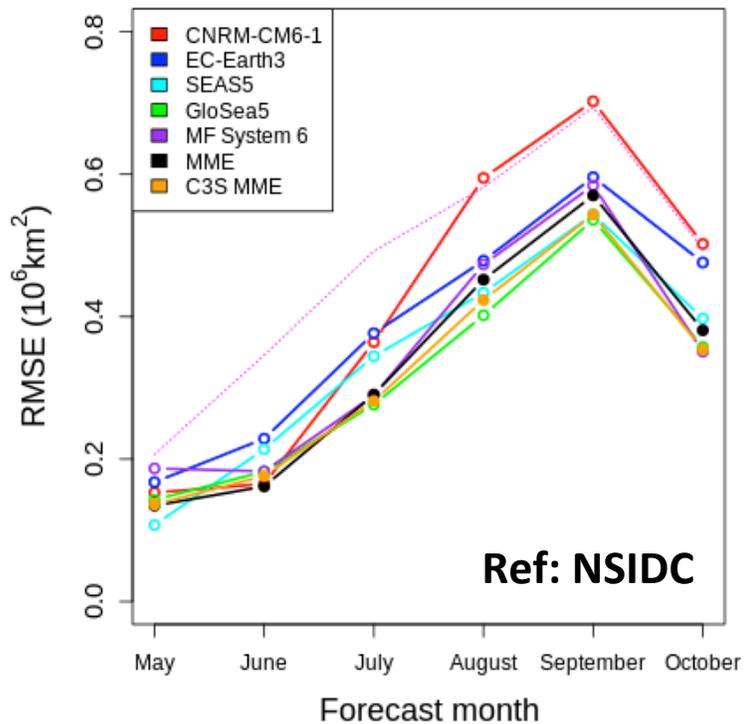
**Task 5.5:** Recommendations for future forecasting system development



# WP5: Stream 1 baseline skill

## STREAM 1 + C3S systems

RMSE of total Arctic SIE



Integrated Ice Edge Error



Goessling et al. 2016



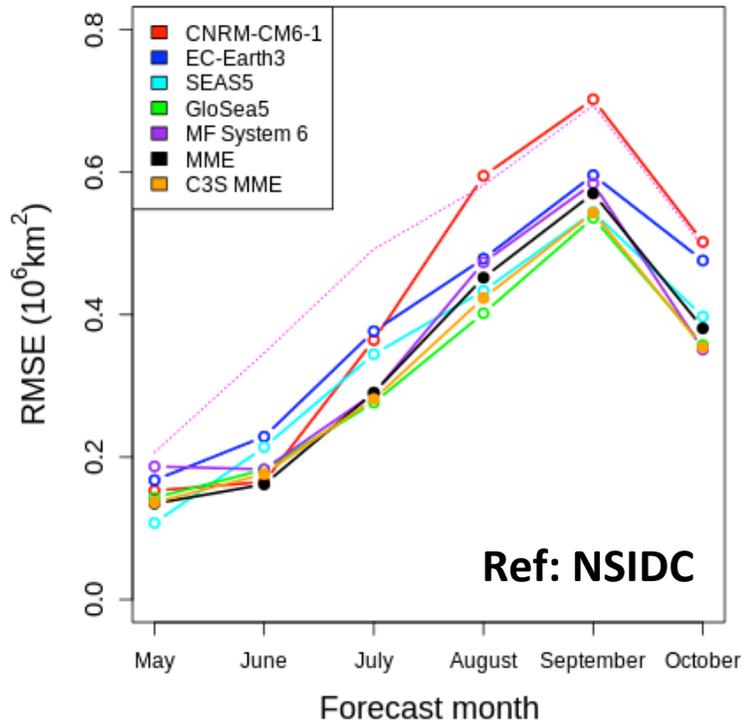
**1st May Initialized Forecasts (1993-2014)**



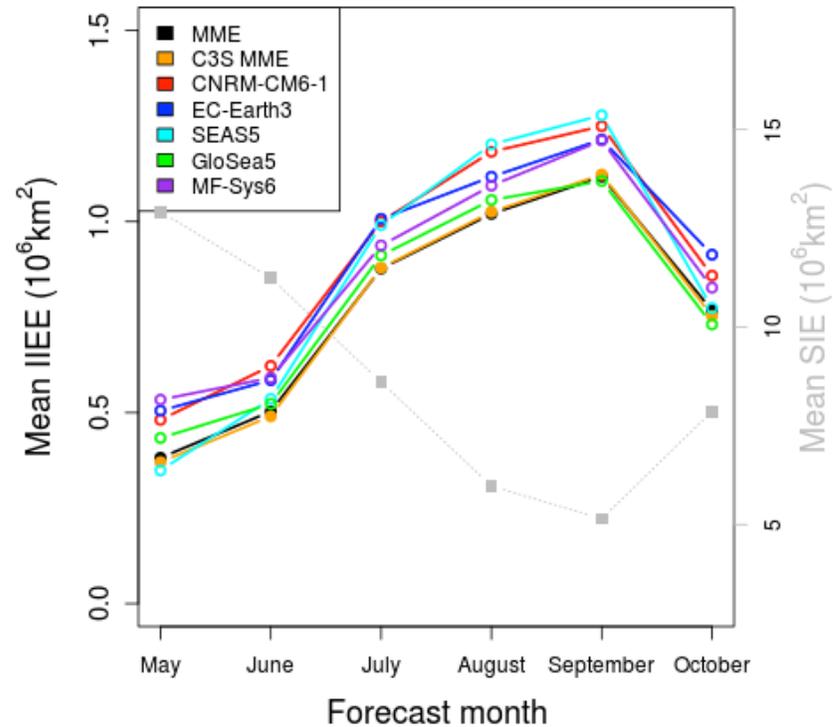
# WP5: Stream 1 baseline skill

## STREAM 1 + C3S systems

RMSE of total Arctic SIE



Pan-Arctic IIEE



Inter-model differences are smaller in IIEE

Multi-models are better than individual models



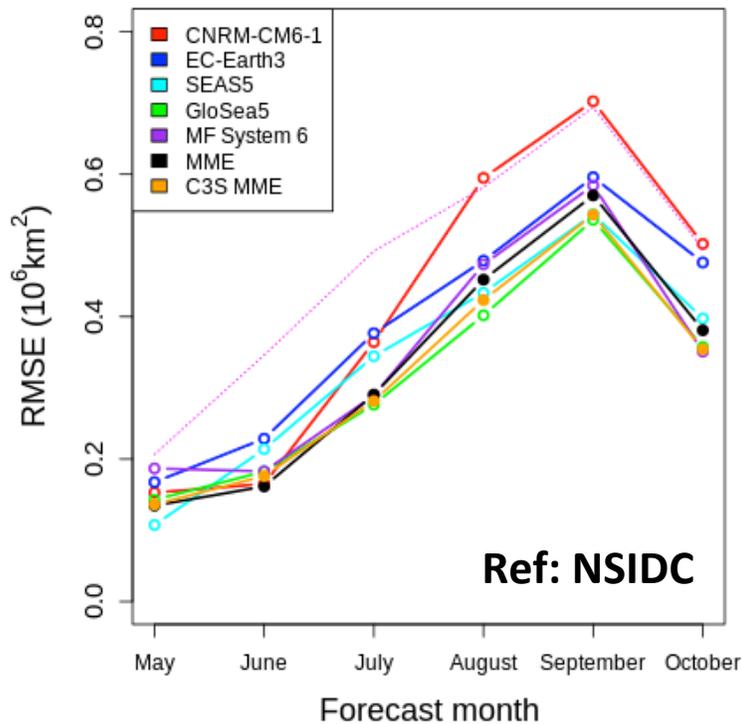
1st May Initialized Forecasts (1993-2014)



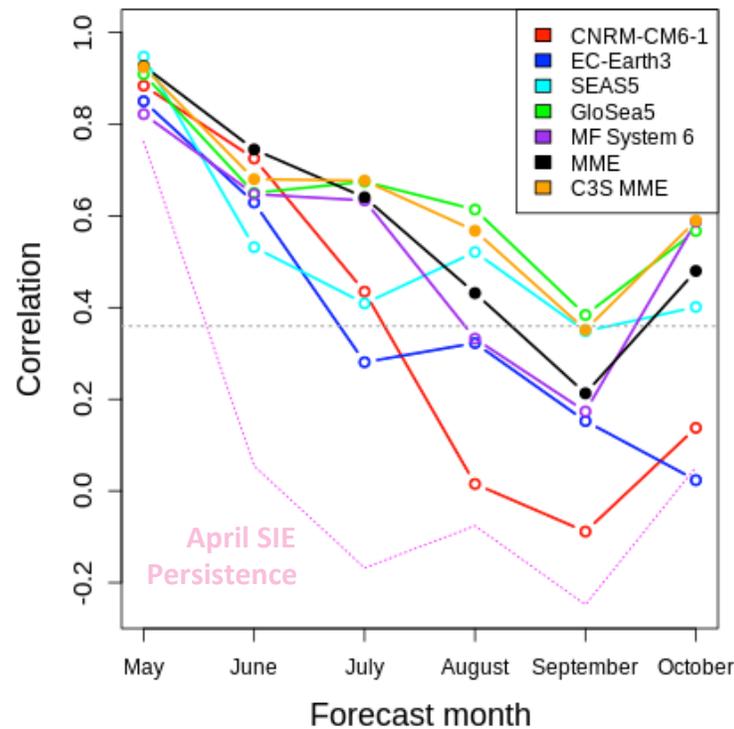
# WP5: Stream 1 baseline skill

## STREAM 1 + C3S systems

RMSE of total Arctic SIE



ACC of total Arctic SIE



Most models exhibit skill up to 3-4 months lead

Forecasting September SIE minimum is still challenging



1st May Initialized Forecasts (1993-2014)



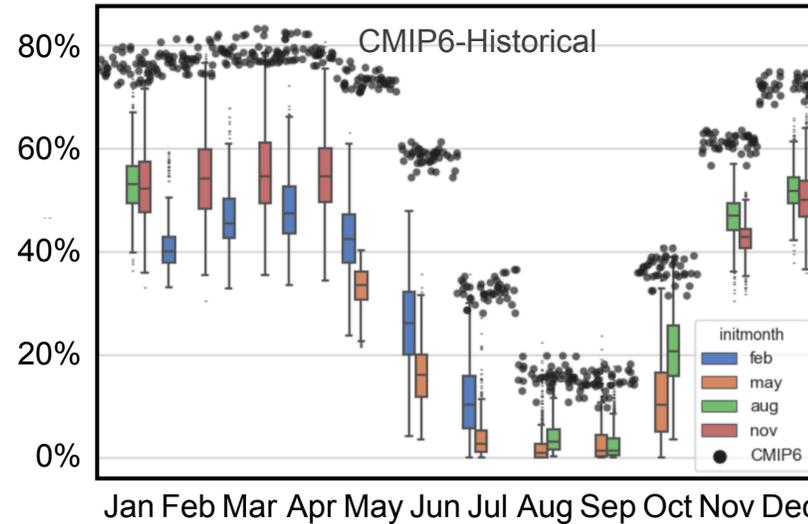
# WP5: Role of Initialization

**Ilona Valisuo**

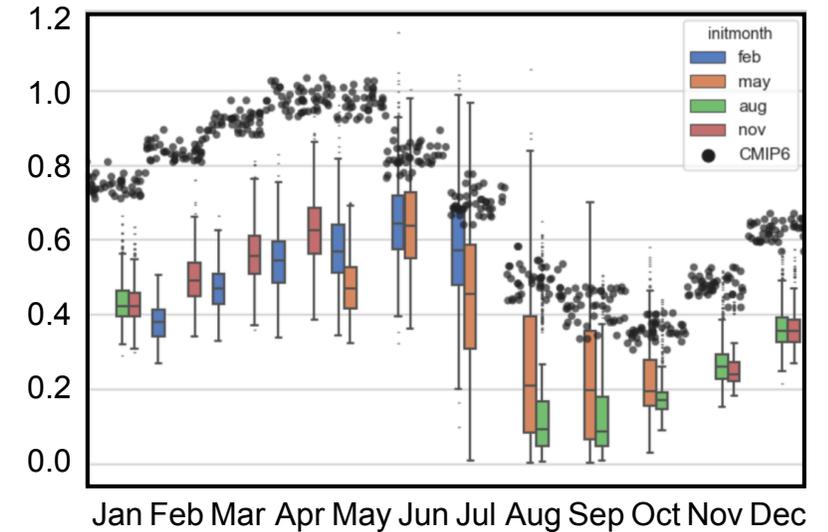
## Greenland Sea



### Sea Ice Concentrations



### Sea Ice Thickness



**CMIP6-Historical has too much and too thick ice in Greenland Sea**  
**Forecasts show huge spread in thickness during the melt season**  
**Systematic error not fully developed by the end of the forecasts**



**CNRM-CM6-1 Seasonal Forecasts (1993-2014)**

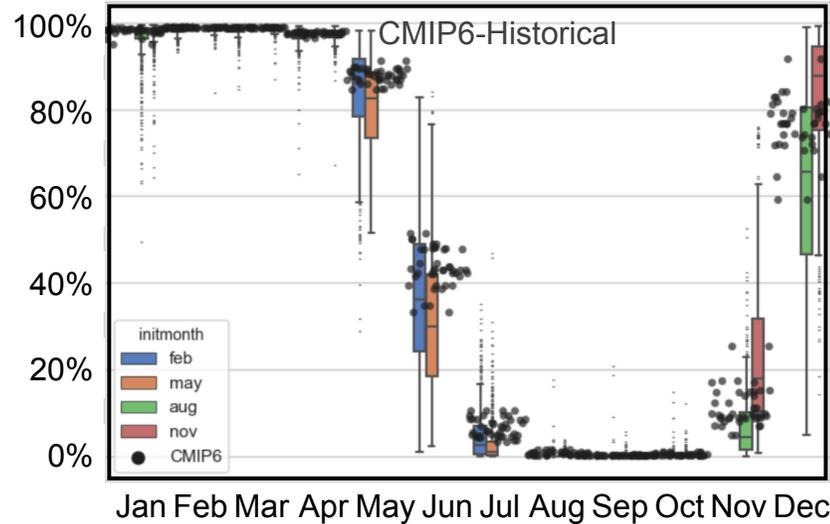


# WP5: Role of Initialization

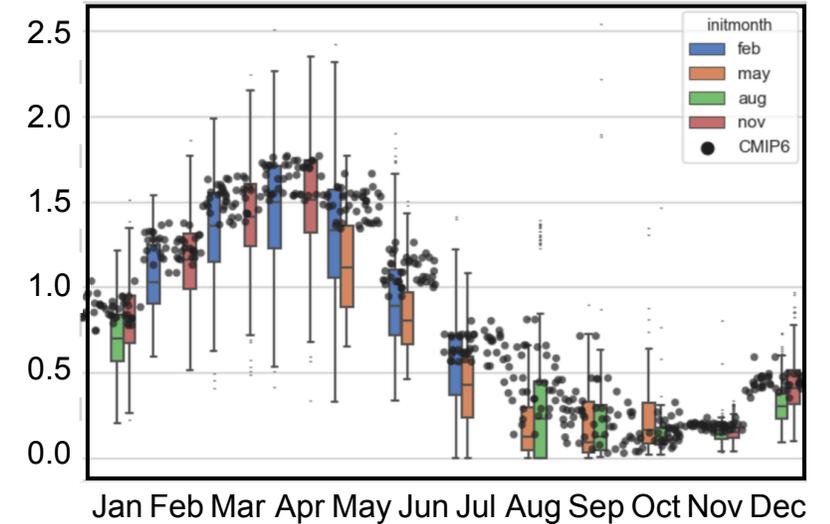
## Chukchi Sea



### Sea Ice Concentrations



### Sea Ice Thickness



**Better agreement between initialized/non initialized forecasts**

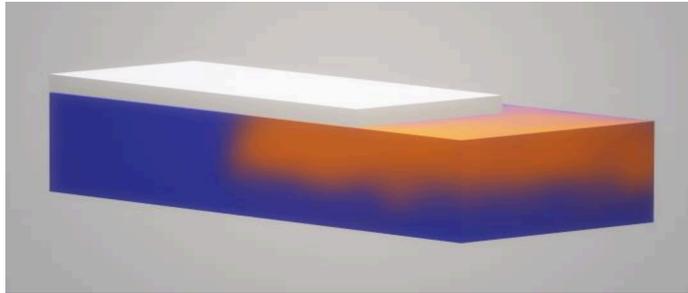


**CNRM-CM6-1 Seasonal Forecasts (1993-2014)**



# WP5: Development of forecast errors

## Inconsistency of ICs



## Initialization Strategy

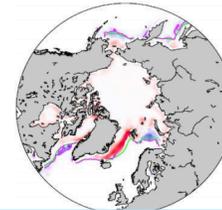
**Sea Ice:** NEMO-LIM3 forced w. ERA-Interim  
ENKF assimilation of SICs from ESA

**Ocean:** ORAS4

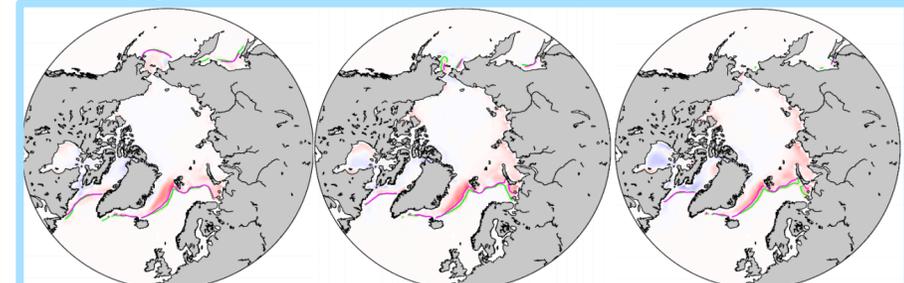
**Atmos:** ERA-Interim

## Evolution of errors with forecast day

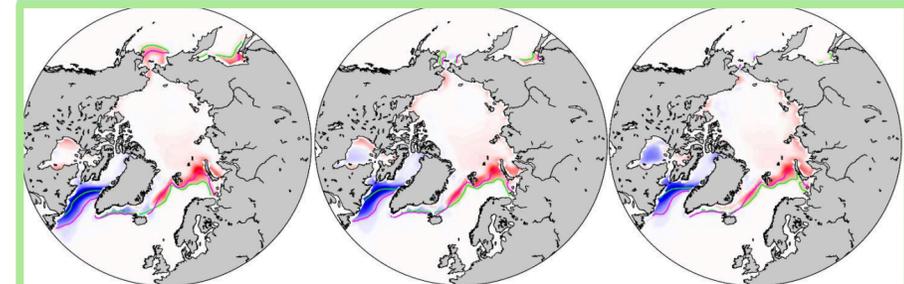
Initial  
Inconsistency



Initialized  
Forecast



Non-initialized  
forecasts



May 10th

May 20th

May 30th

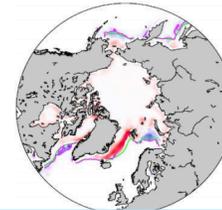
Ice concentration (%)



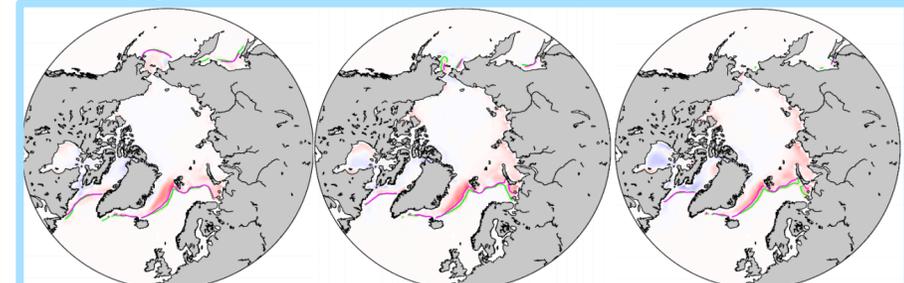
# WP5: Development of forecast errors

## Evolution of errors with forecast day

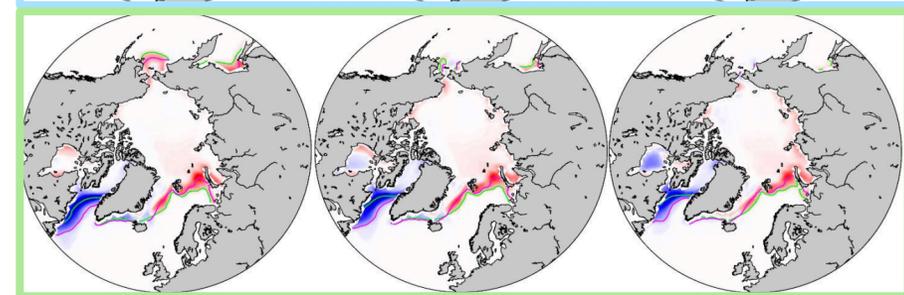
Initial  
Inconsistency



Initialized  
Forecast



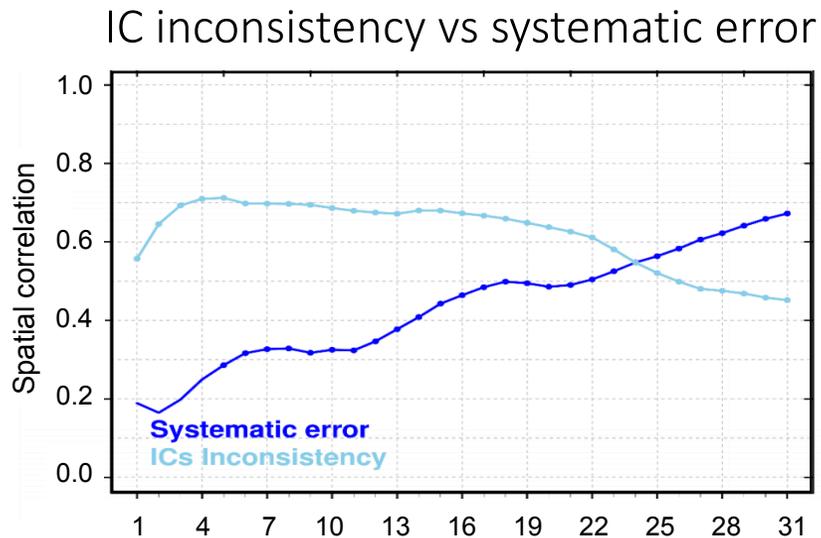
Non-initialized  
forecasts



May 10th

May 20th

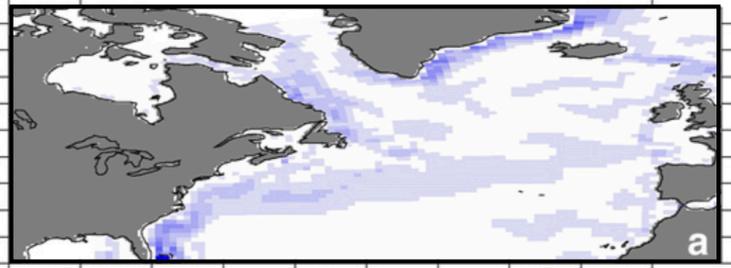
May 30th



# WP5: Impact of model resolution

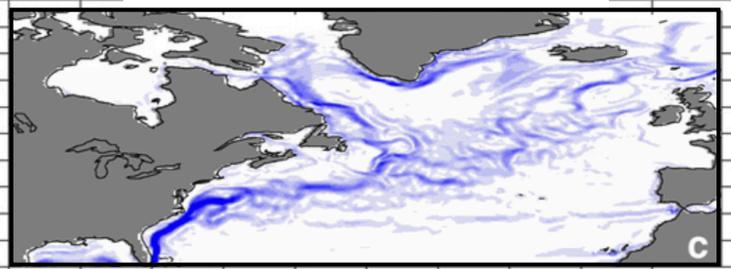
**Juan C. Acosta-Navarro**

NEMO/LIM3 (ORCA1)



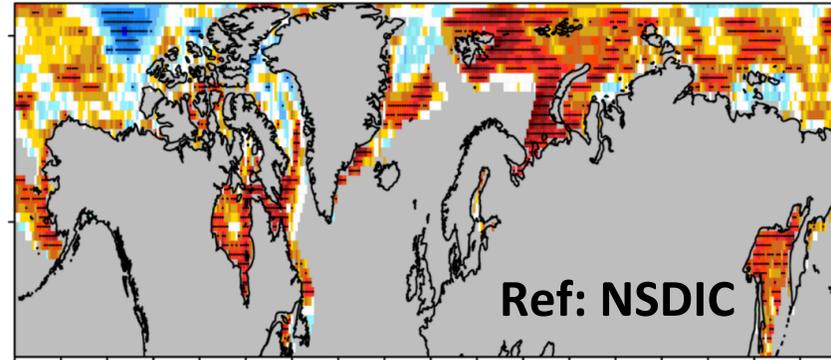
Ocean ICs: forced run nudged to ORAS4

NEMO/LIM3 (ORCA025)

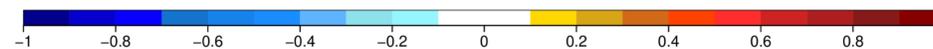
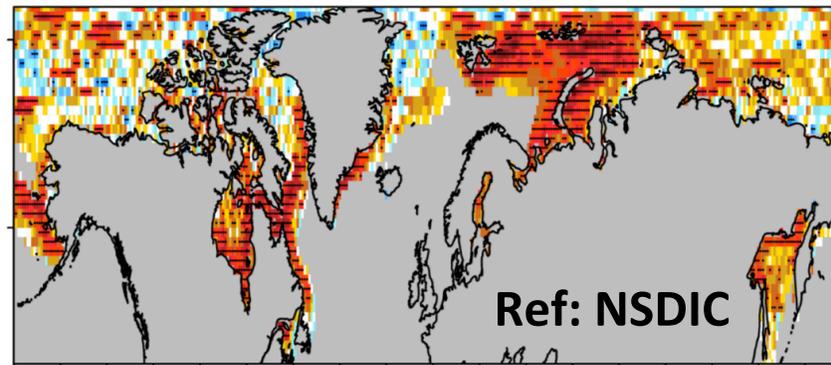


Ocean ICs: forced run nudged to ORAS5

ACC in Sea Ice Concentration in DJF (ORCA1)



ACC in Sea Ice Concentration in DJF (ORCA025)

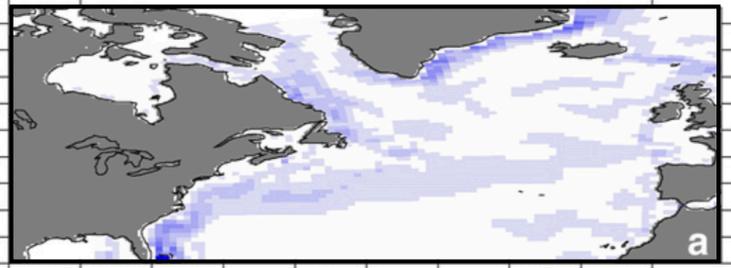


**EC-Earth3.3** November Forecasts (1993-2014)



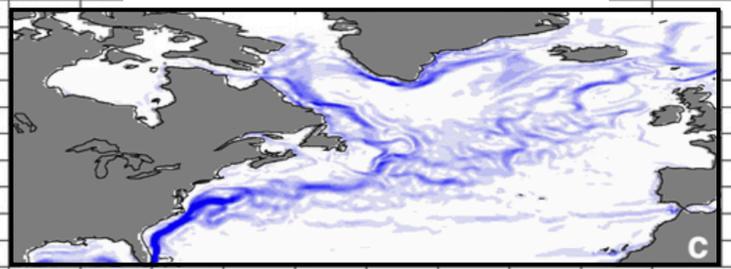
# WP5: Impact of model resolution

NEMO/LIM3 (ORCA1)



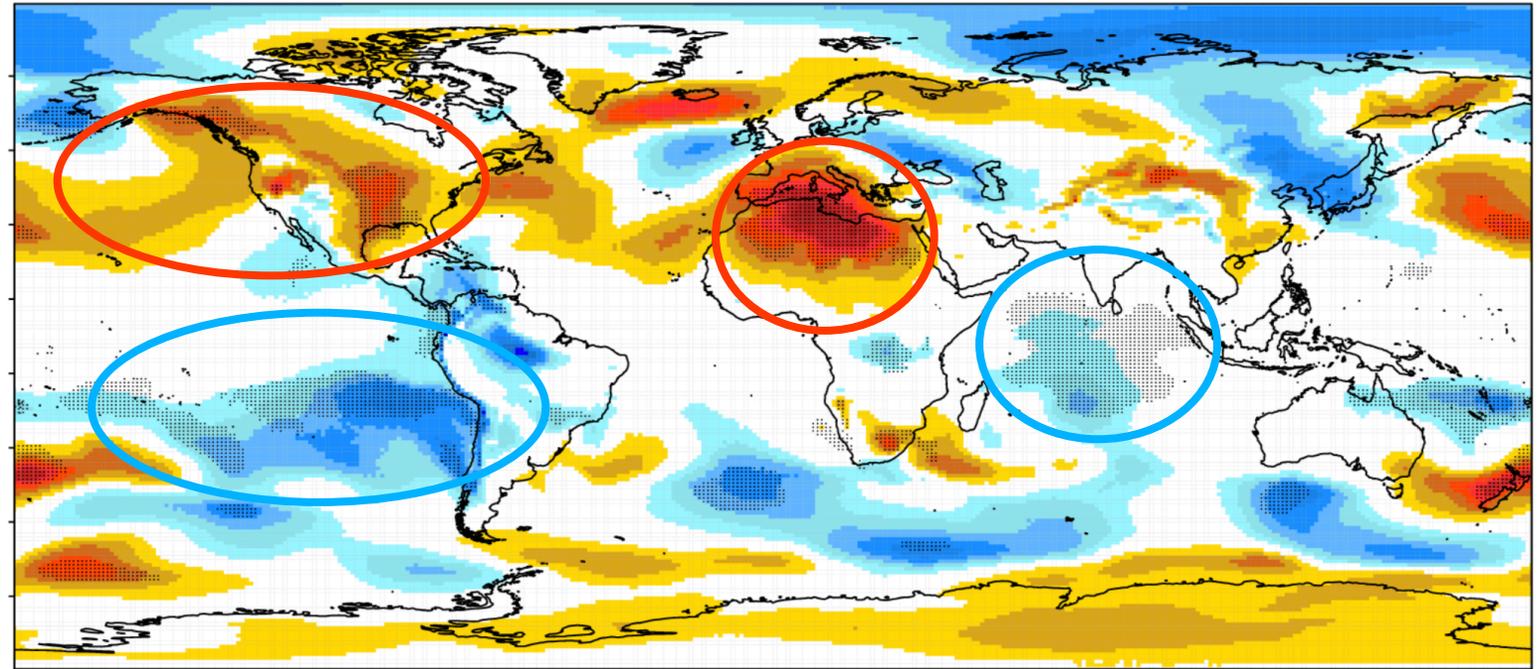
Ocean ICs: forced run nudged to ORAS4

NEMO/LIM3 (ORCA025)



Ocean ICs: forced run nudged to ORAS5

ACC difference in DJF SLP (ORCA025 – ORCA1)

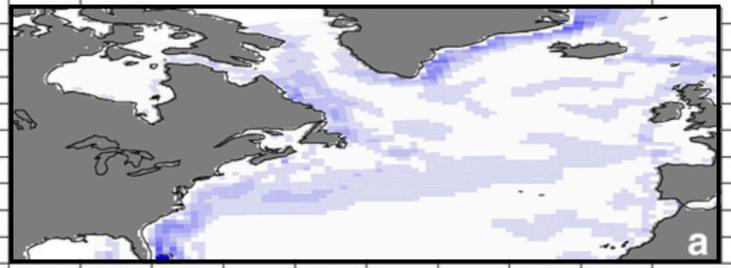


Ref: ERA-Interim



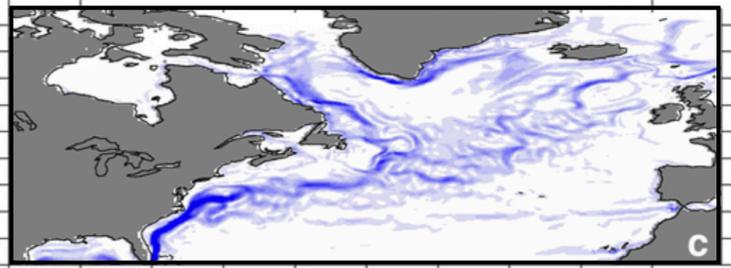
# WP5: Impact of model resolution

NEMO/LIM3 (ORCA1)



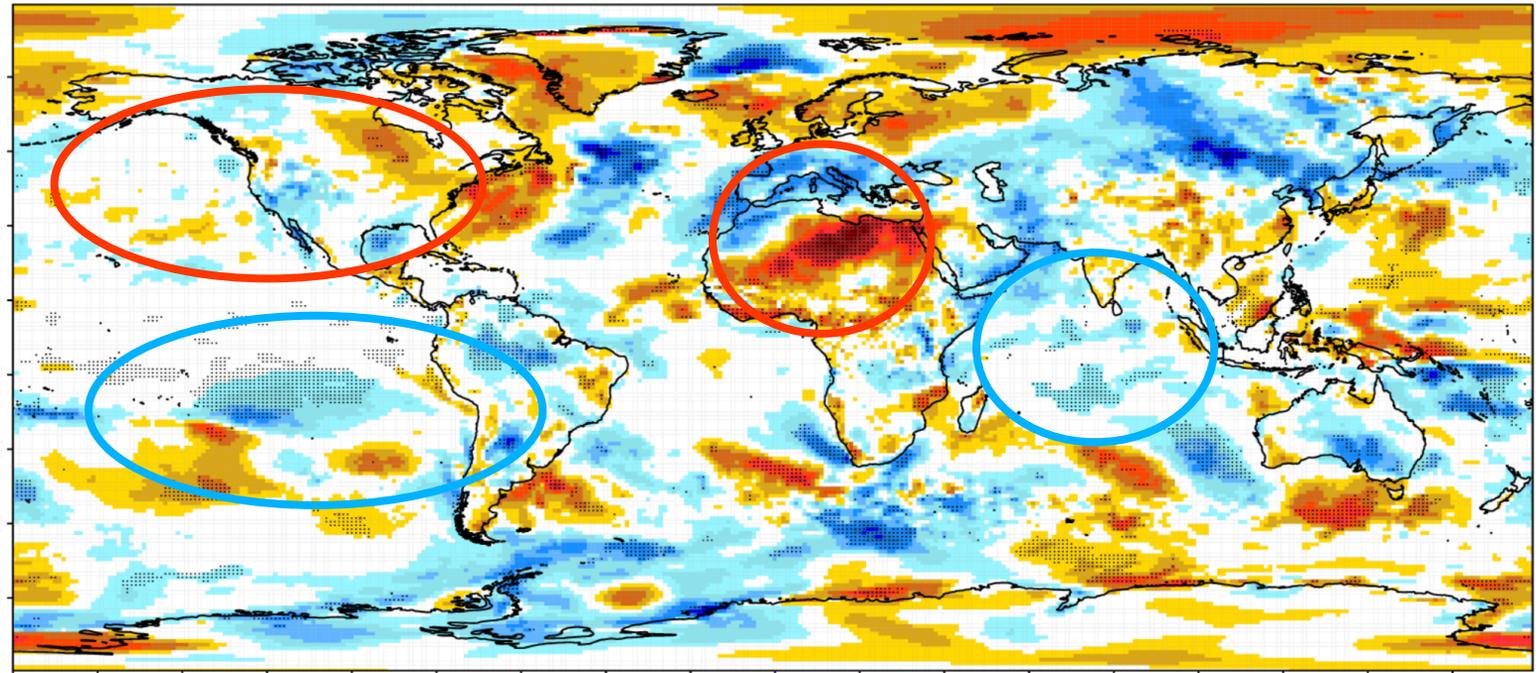
Ocean ICs: forced run nudged to ORAS4

NEMO/LIM3 (ORCA025)



Ocean ICs: forced run nudged to ORAS5

ACC difference in DJF TAS (ORCA025 – ORCA1)

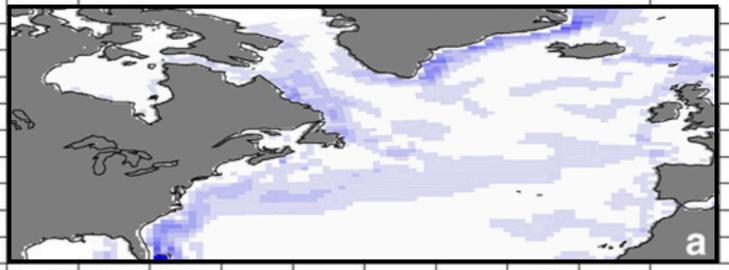


Ref: ERA-Interim



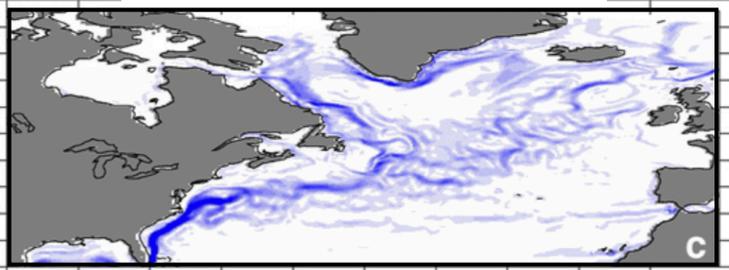
# WP5: Impact of model resolution

NEMO/LIM3 (ORCA1)



Ocean ICs: forced run nudged to ORAS4

NEMO/LIM3 (ORCA025)



Ocean ICs: forced run nudged to ORAS5

## Systematic Analysis with three GCMs



**EC-Earth3.3** November Forecasts (1993-2014)



# WP5: Statistical climate predictions

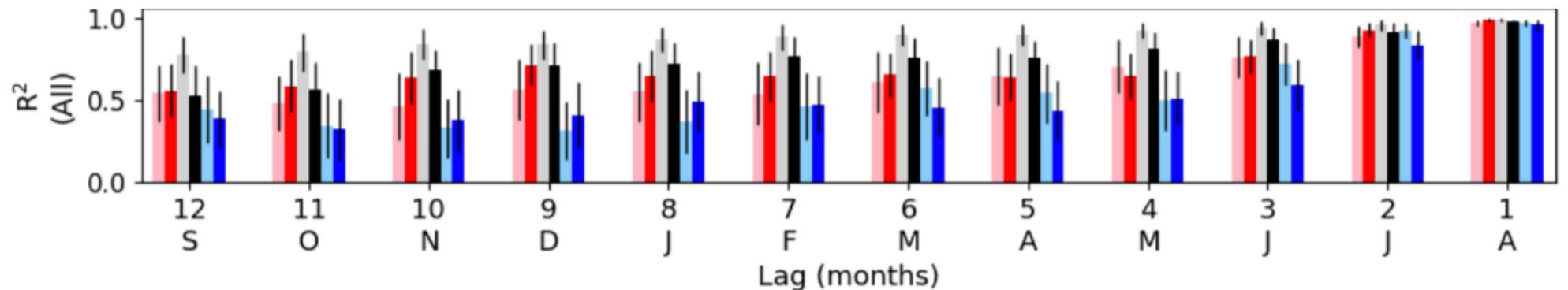
## 7 different predictors

SIV: Sea Ice Volume  
OHT: Ocean Heat Transport  
SIC: Sea Ice Concentration  
SID: Sea Ice Drift  
SIA: Sea Ice Area  
SIT: Sea Ice Thickness  
SST: Sea Surface Temperature

## Outputs from 6 models

HadGEM3-LL  
HadGEM3-MM  
ECMWF-LR  
ECMWF-HR  
AWI-LR  
AWI-HR

Statistical predictability of September SIV Anomaly for 1 to 12 preceding months



# WP5: Optimal sampling locations

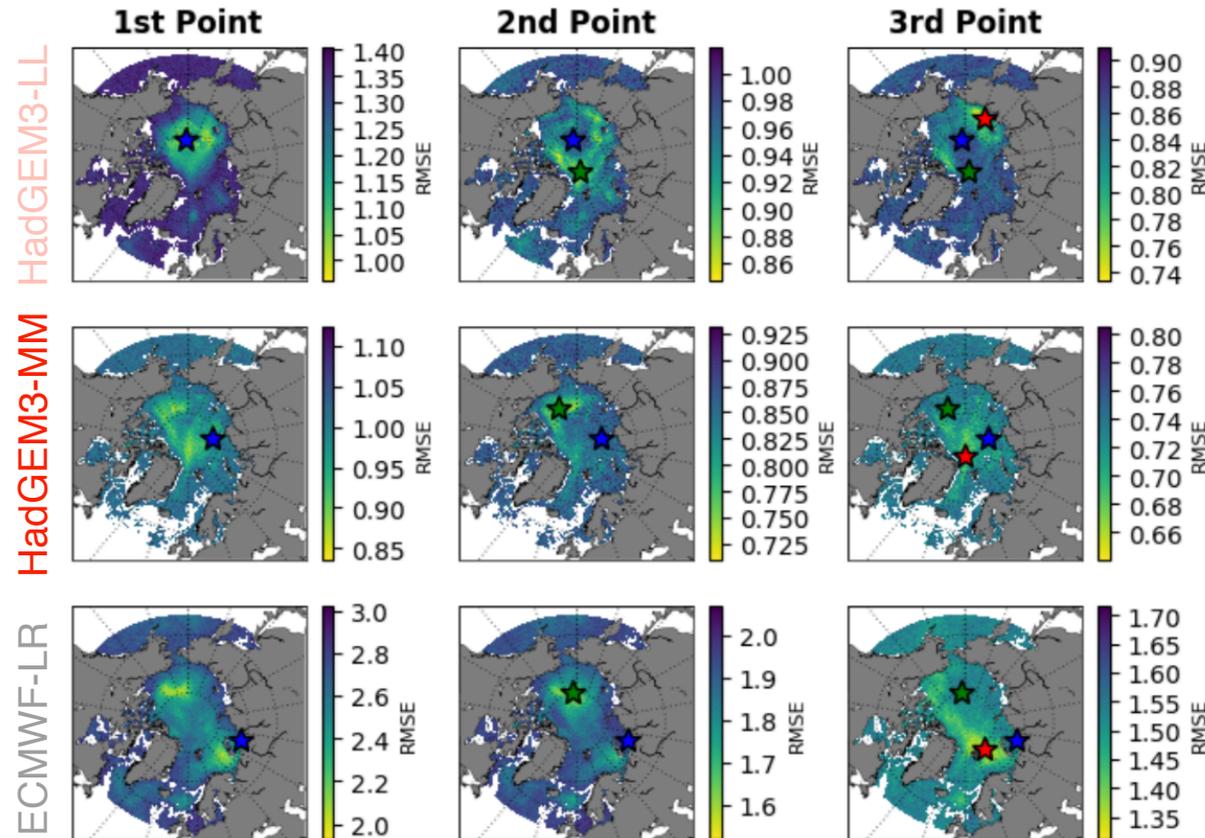
## 4 predictors (easily observable)

Sea Ice Drift (in-situ)  
Sea Ice Thickness (in-situ)  
Sea Ice Concentration (satellite)  
Sea Surface Temperature (in-situ)

## Optimal locations:

Placed at the grid points where  
predictors minimise RMSE in SIV

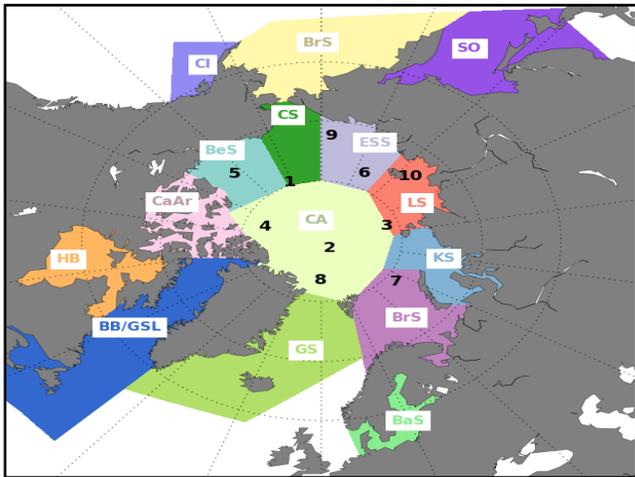
## RMSE of Sea Ice Volume



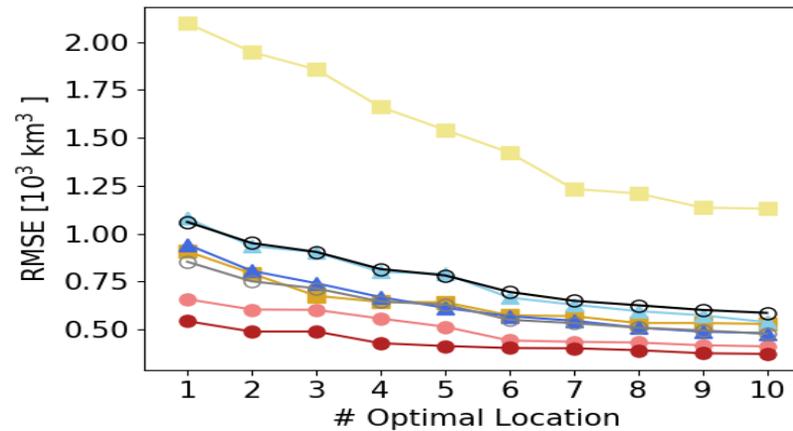
# WP5: Optimal sampling locations

## 4 predictors (easily observable)

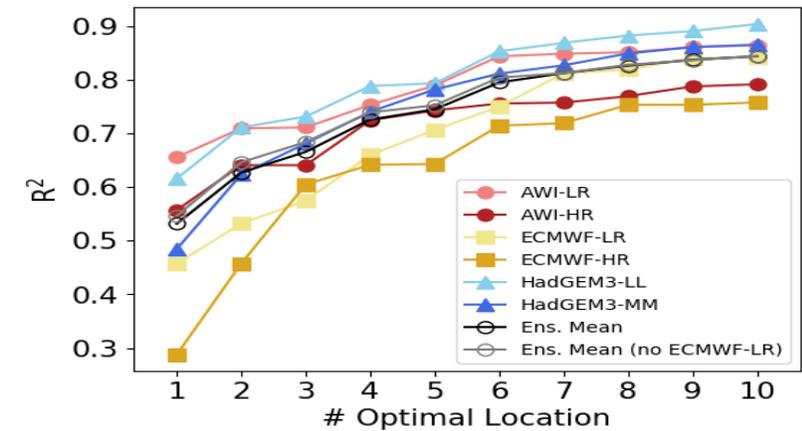
Sea Ice Drift (in-situ)  
Sea Ice Thickness (in-situ)  
Sea Ice Concentration (satellite)  
Sea Surface Temperature (in-situ)



RMSE of Total SIV



R<sup>2</sup> of Total SIV



**5 to 6 locations can guarantee a relatively low RMSE and high R**  
**ECMWF-LR has a strong RMSE bias, which creates too thick sea ice**



- Advances predictive capacity in polar regions and beyond:
  - Develops models with enhanced representation of Arctic processes
  - Contributes to improving the Arctic observing system
- Enhances our understanding of Arctic-midlatitude linkages (also from a prediction perspective)
- Brings different communities closer together



# Highlights of APPLICATE WP5

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Experimental framework to foster the predictive skill over the Arctic

- APPLICATE Stream 1 seasonal forecasts show **skill to predict summer SIE up to 3-4 months** beforehand
- Increasing the resolution seems to lead to higher predictive skill in the Northern Hemisphere, although it is unclear if the improvement comes from the ICs or from the resolution itself
- Statistical models can achieve high level of skill up to 12 months ahead



Thanks for your attention!!



For further questions, please contact me at: [pablo.ortega@bsc.es](mailto:pablo.ortega@bsc.es)