JLESC Research Collaboration: Monitoring the Arctic Climate



Data Assimilation Research Team, RIKEN Center for Computational Science

T. Miyoshi, S. Kotsuki, K. Terasaki



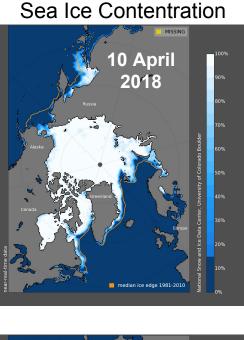
Earth Science Department

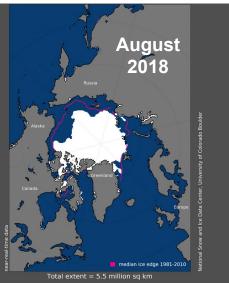
V. Guemas, K.Serradell, <u>P. Ortega</u>,

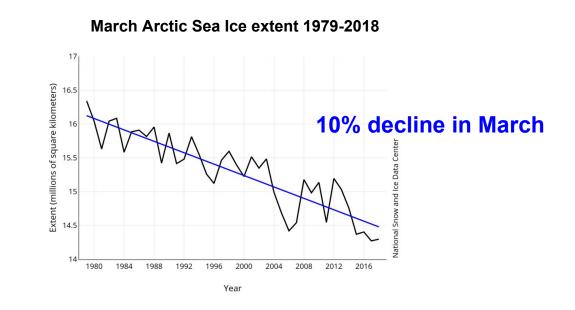
- M. Castrillo, M. Acosta, J. Acosta,
- P. Echevarria, E. Moreno-Chamarro

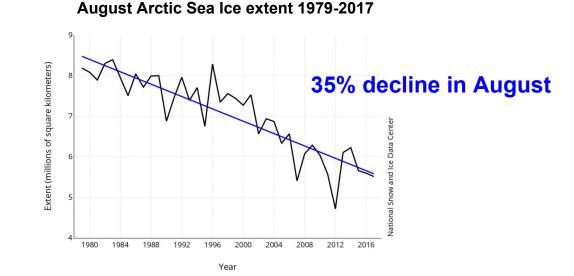
Context: A dramatic decline in Arctic sea ice...

Near-real time satellite observations from NSIDC

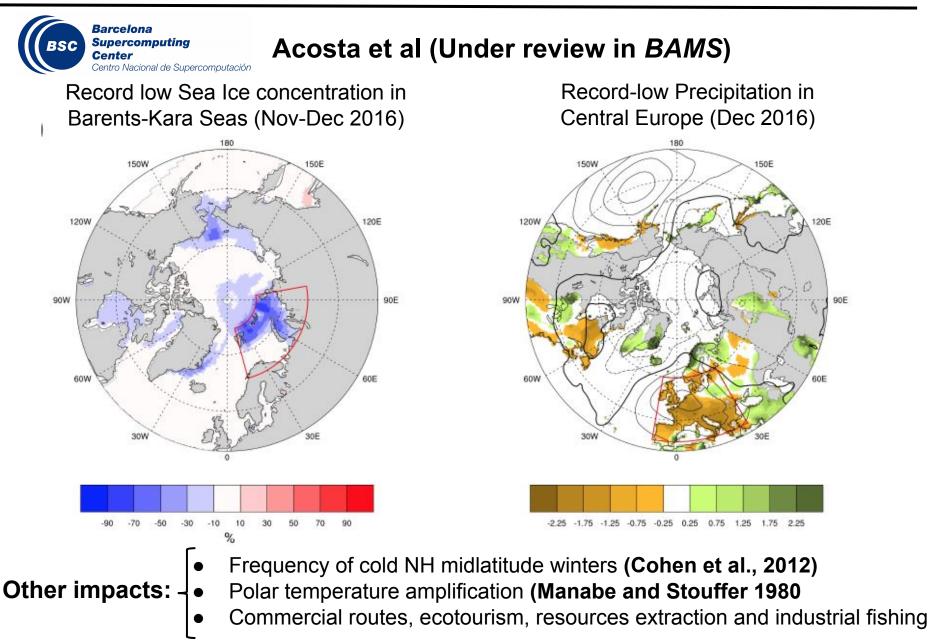




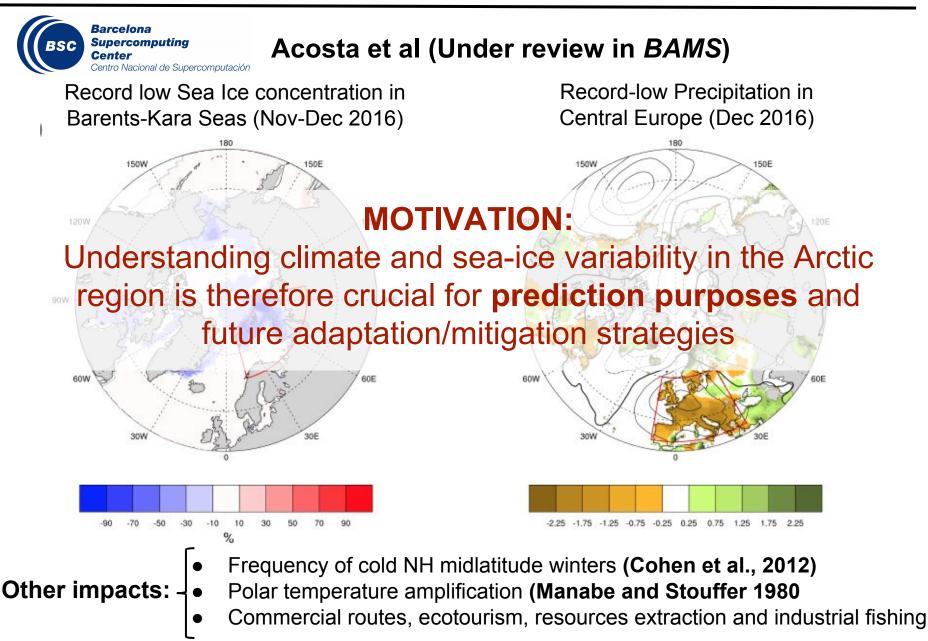




Context: ...with important climatic consequences



Context: ...with important climatic consequences



Cornerstones of Numerical Weather Prediction

For a given

atmospheric state

AEMet

EUMETSAT

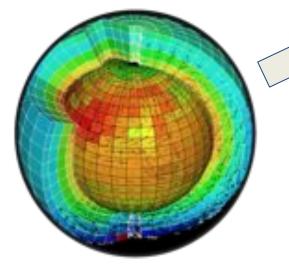
Agencia Estatal de Meteorologí

we can solve the Governing equations

 $\rho_{o}D_{t}\vec{v} + 2\Omega \times \rho_{o}\vec{v} + g\rho\hat{k} + \nabla p = \vec{F}$ $\rho_{o}\nabla \cdot \vec{v} = 0$ $\partial_{t}\eta + \nabla \cdot (H + \eta)\vec{v}_{h} = P - E$ $D_{t}\theta = Q_{\theta}$ $D_{t}s = Q_{s}$ $\rho = \rho(s,\theta,p)$

To make a weather prediction

within a discretized global climate model

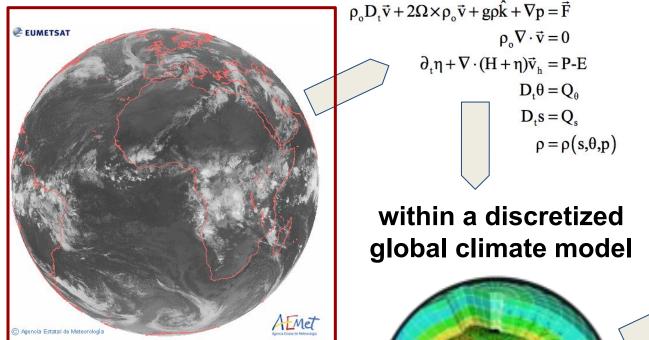


from: https://www.earthsystemcog.org/ projects/esmf/

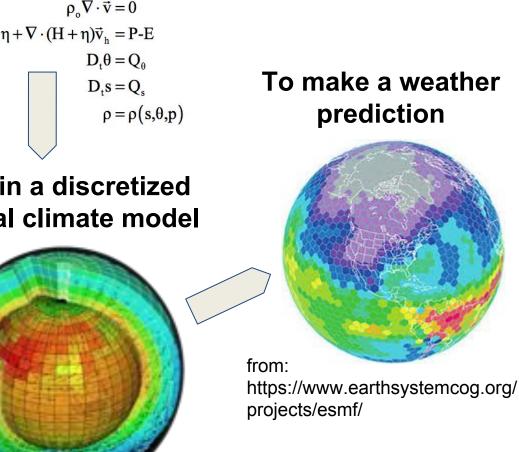
Cornerstones of Numerical Weather Prediction

For a given atmospheric state

we can solve the Governing equations



Good predictions rely on a good initialization of the model with observational data



NICAM-LETKF: Global Atmospheric Data Assimilation System

S. Kotsuki, K. Terasaki, T. Miyoshi

Data Assimilation Research Team, RIKEN Center for Computational Science, Japan





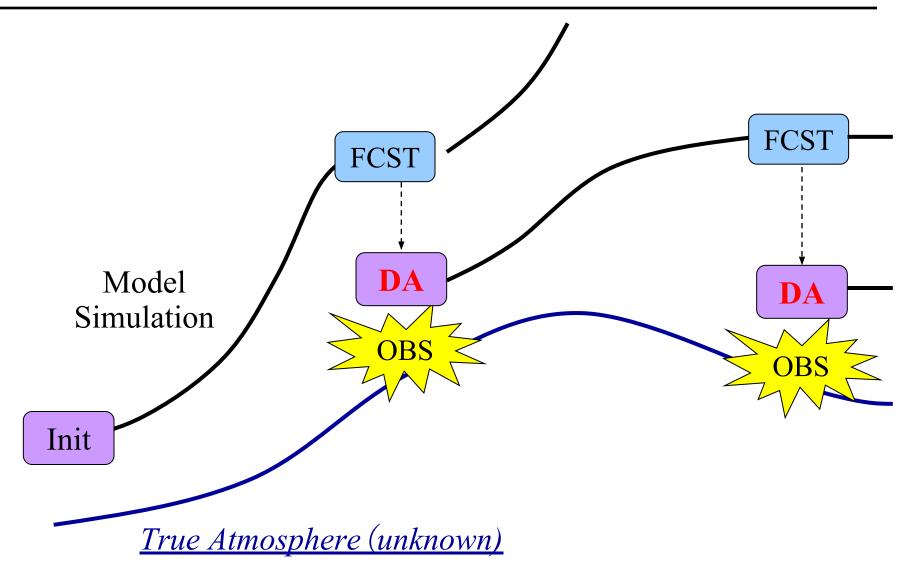
Overview on Data Assimilation



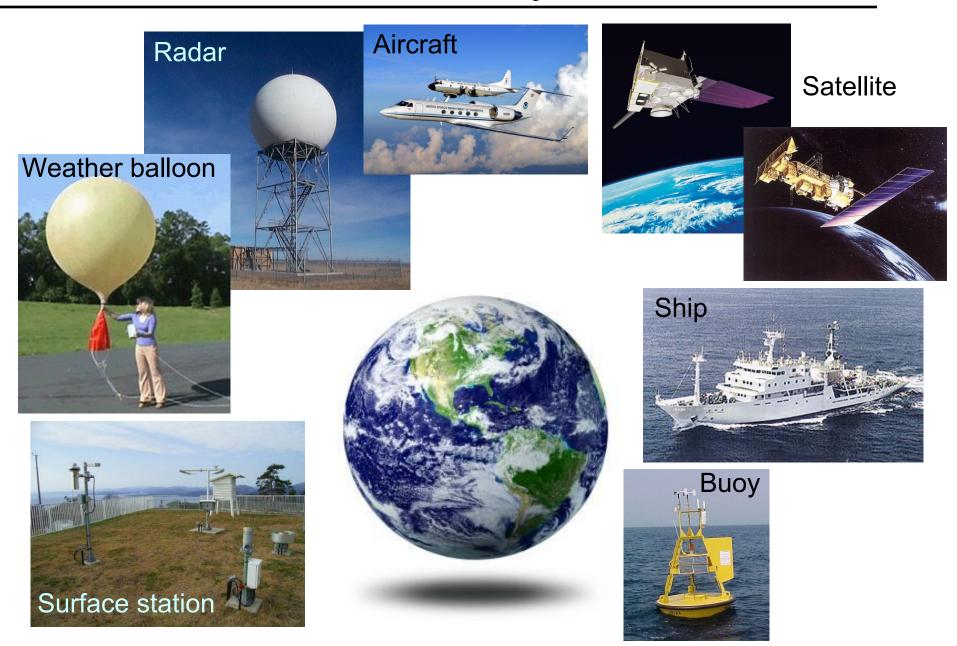
Local Ensemble Transform Kalman Filter *(Hunt et al. 2007)*

Data assimilation combines simulation and real world, and brings synergy

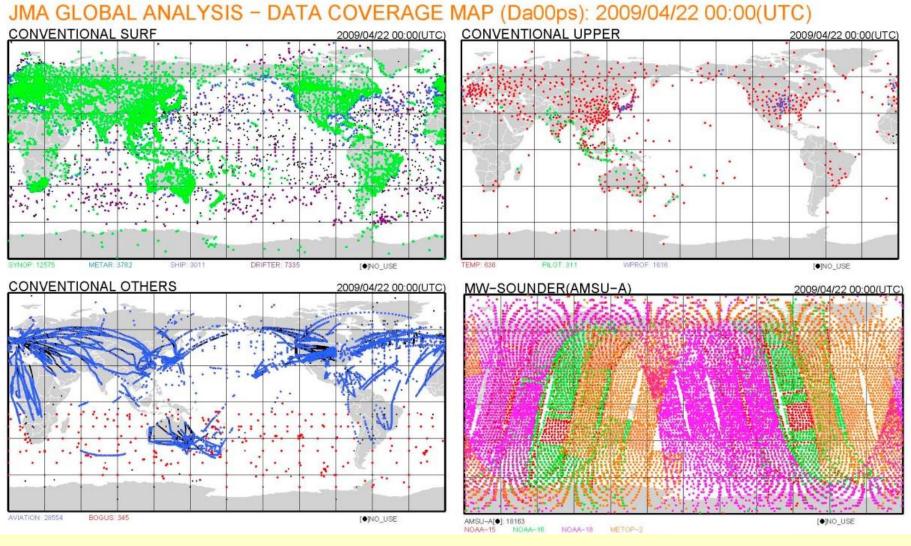
Numerical Weather Prediction



Global Observation System

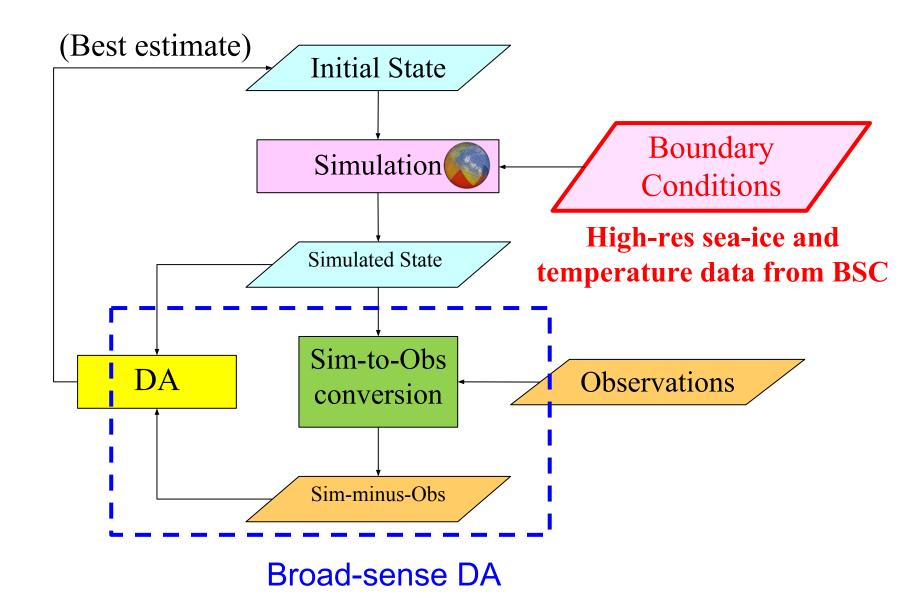


Observation data (6-h period) (Courtesy of JMA)

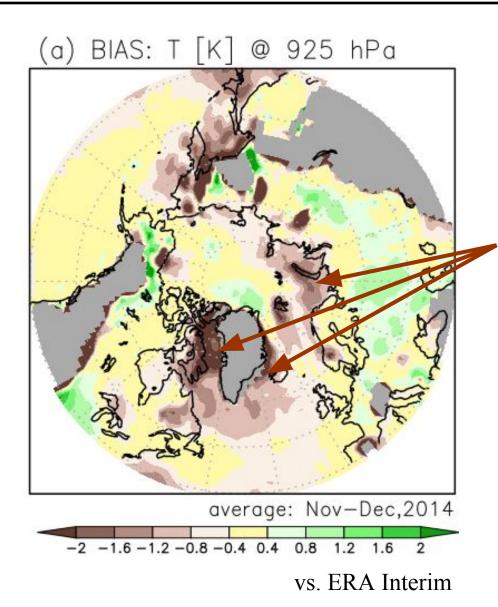


NWP has been pioneering "Big Data" science!

Workflow of Data Assimilation



Caveat: Temperature Bias in NICAM-LETKF

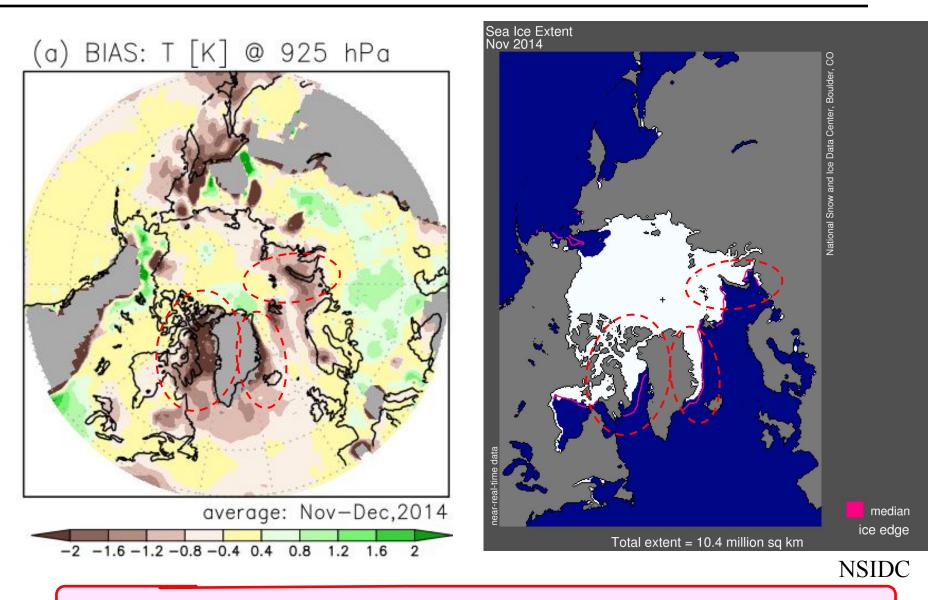


Present NICAM-LETKF has

a cold BIAS

near the surface.

Caveat: Temperature Bias in NICAM-LETKF



Sea ice and temperature data may mitigate this problem!

www.bsc.es

Earth Sciences Department



Barcelona Supercomputing Center Centro Nacional de Supercomputación



Arctic research activities at BSC:

From Sea Ice reconstructions to Seasonal-to-decadal (S2D) Prediction

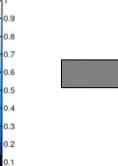
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Arctic Research at BSC

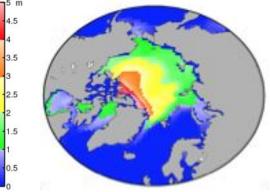
- ICs for S2D Climate Predictions Sea Ice Reconstruction:
 - Provides smoother initialization than observational products
 - \circ Covers the period 1979-2015
 - Performed with coupled version of NEMO+LIM in Standard resolution (1°x1°)
 - Boundary conditions: Atmospheric forcing from DFS5.2
 - Assimilation of ESA and OSISAF (two european satellite products) Sea Ice concentrations using EnKF

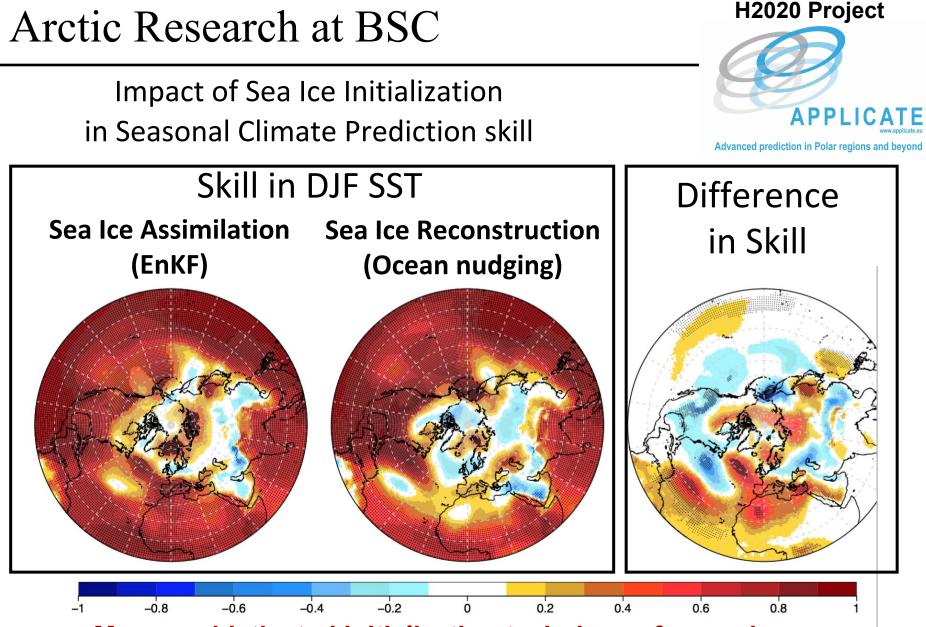
Observed Sea Ice Concentration Sep 2007





Reconstructed Sea Ice Thickness Sep 2007





More sophisticated initialization techniques for sea-ice (EnKF) improve prediction skill in the mediterranean region

Next steps of the Collaboration

- Porting the Ec-Earth Model to the K-Computer
- Developing and testing a new configuration to assimilate sea-ice with EnKF at high-resolution
- Producing a high-resolution sea ice reconstruction
- Incorporating the new HR reconstruction in the NICAM-LETKF Assimilation System
- Investigating the impact of these improved sea-ice initial conditions on the skill of NWPs

Perspective

"Big Data" \Lapha "Big Simulation"

For the next decade: Exa-scale computing

With the Post-K, we aim to run 1000-member global NICAM-LETKF at 3.5-km resolution

→ Revolutionize Weather Prediction

in close collaboration with the FLAGSHIP 2020 project

