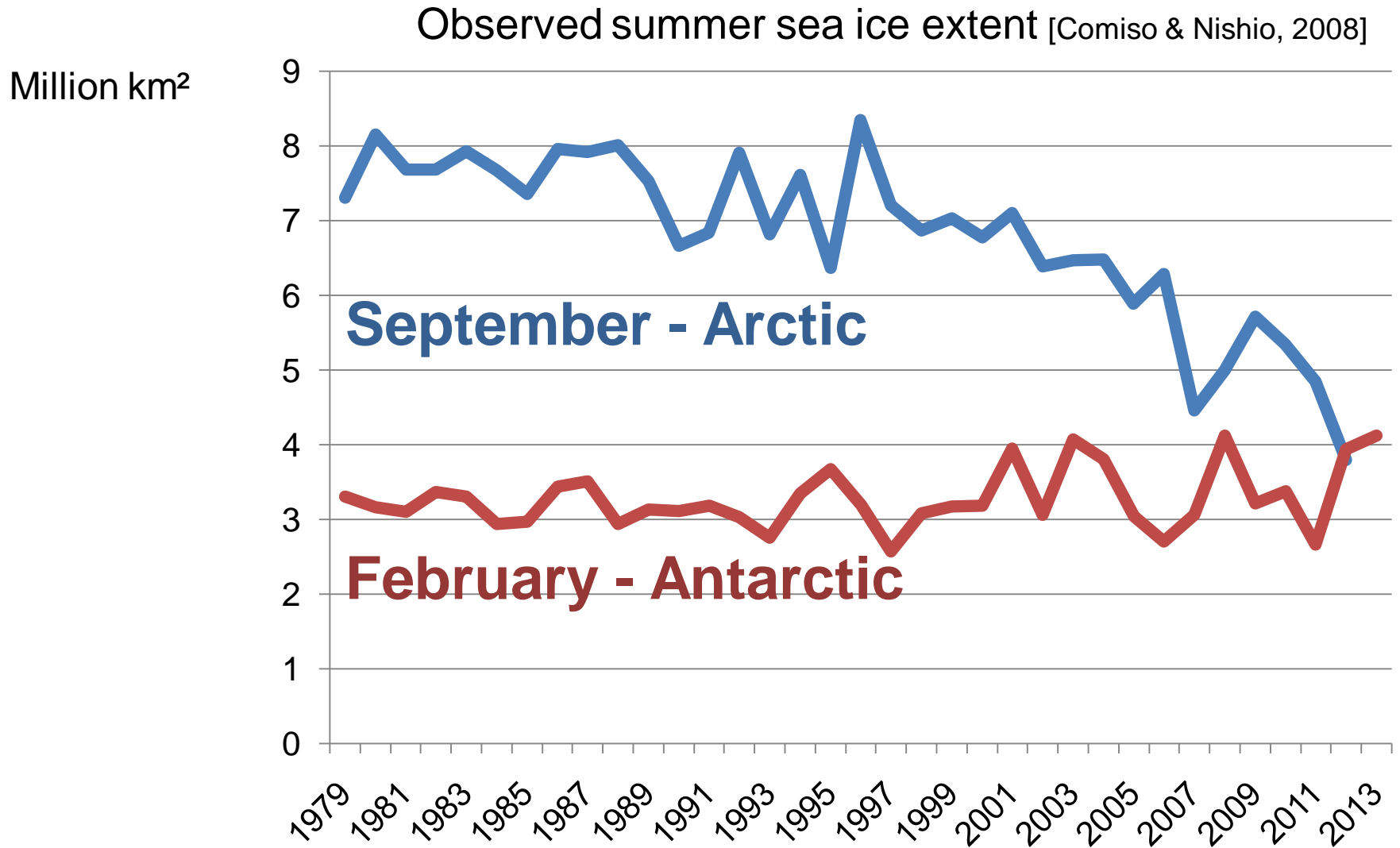


Ensemble Kalman filter in sea ice modeling

François Massonnet
T. Fichefet • H. Goosse



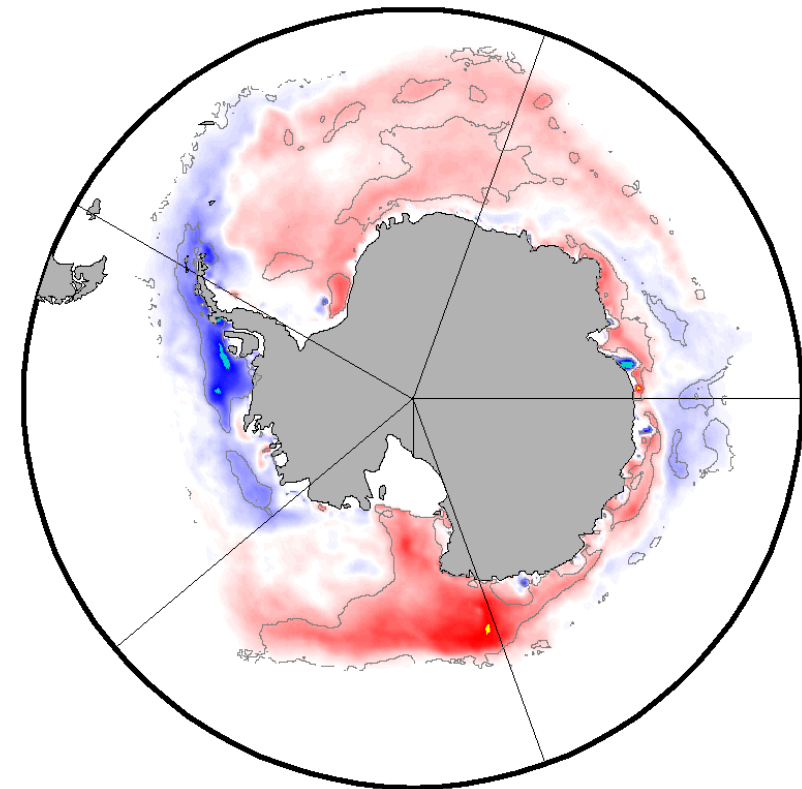
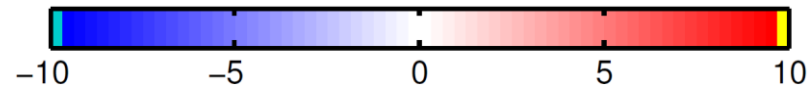
The 2012 sea ice kiss: two pending challenges



Challenge 1

Expanding Antarctic sea ice

Sea ice concentration trend [%/decade]



- Mostly **areal** observations

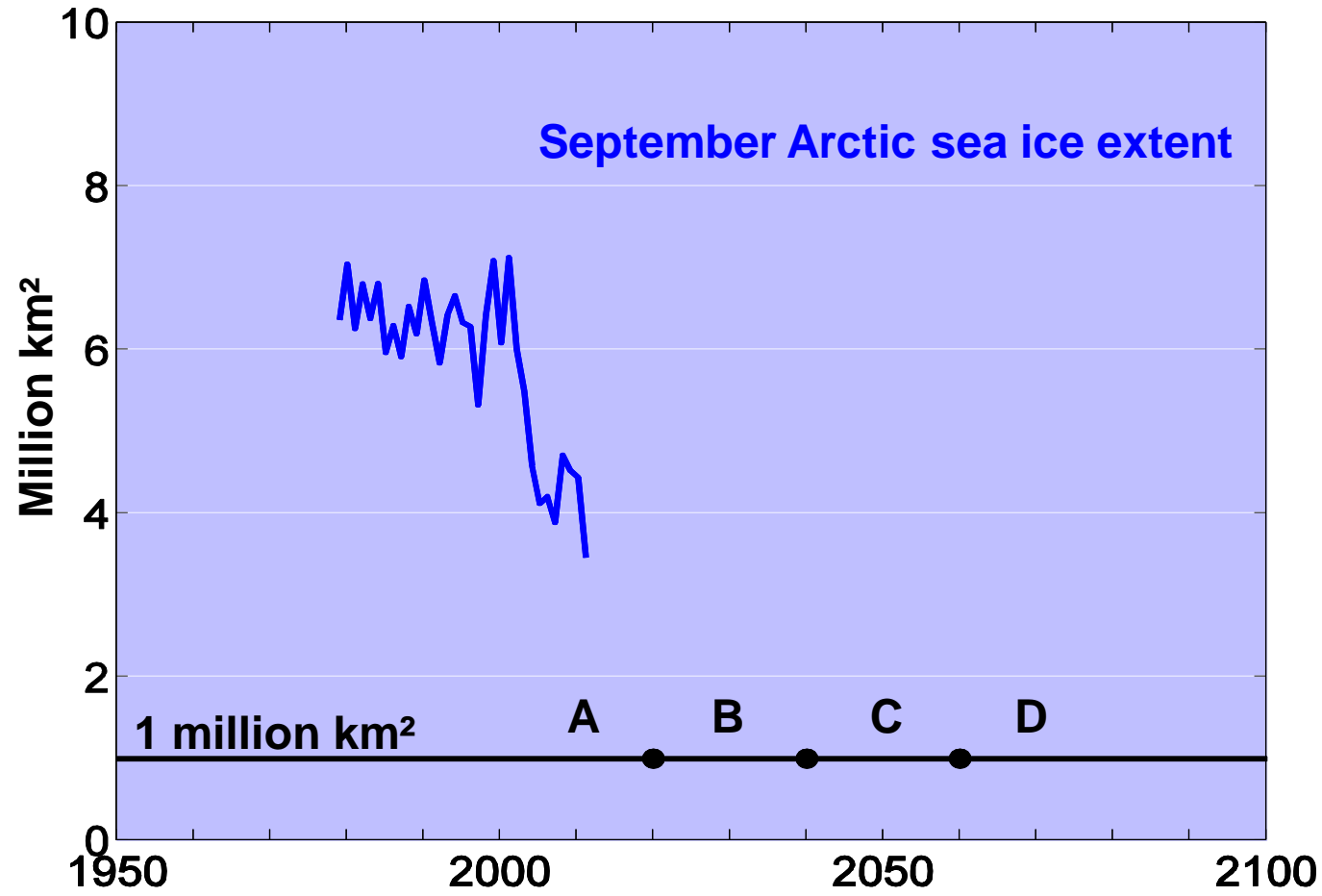
- **GCMs are of limited utility**
because of biased mean state and
variability [Zunz et al., 2013]

- **Several interpretations proposed**

- Changes in winds [Holland and Kwok, 2012]
- Changes in hydrological cycle
[Zhang, 2007; Bintanja et al., 2013]
- Unforced variability [Polvani & Smith, in rev.]
- ...

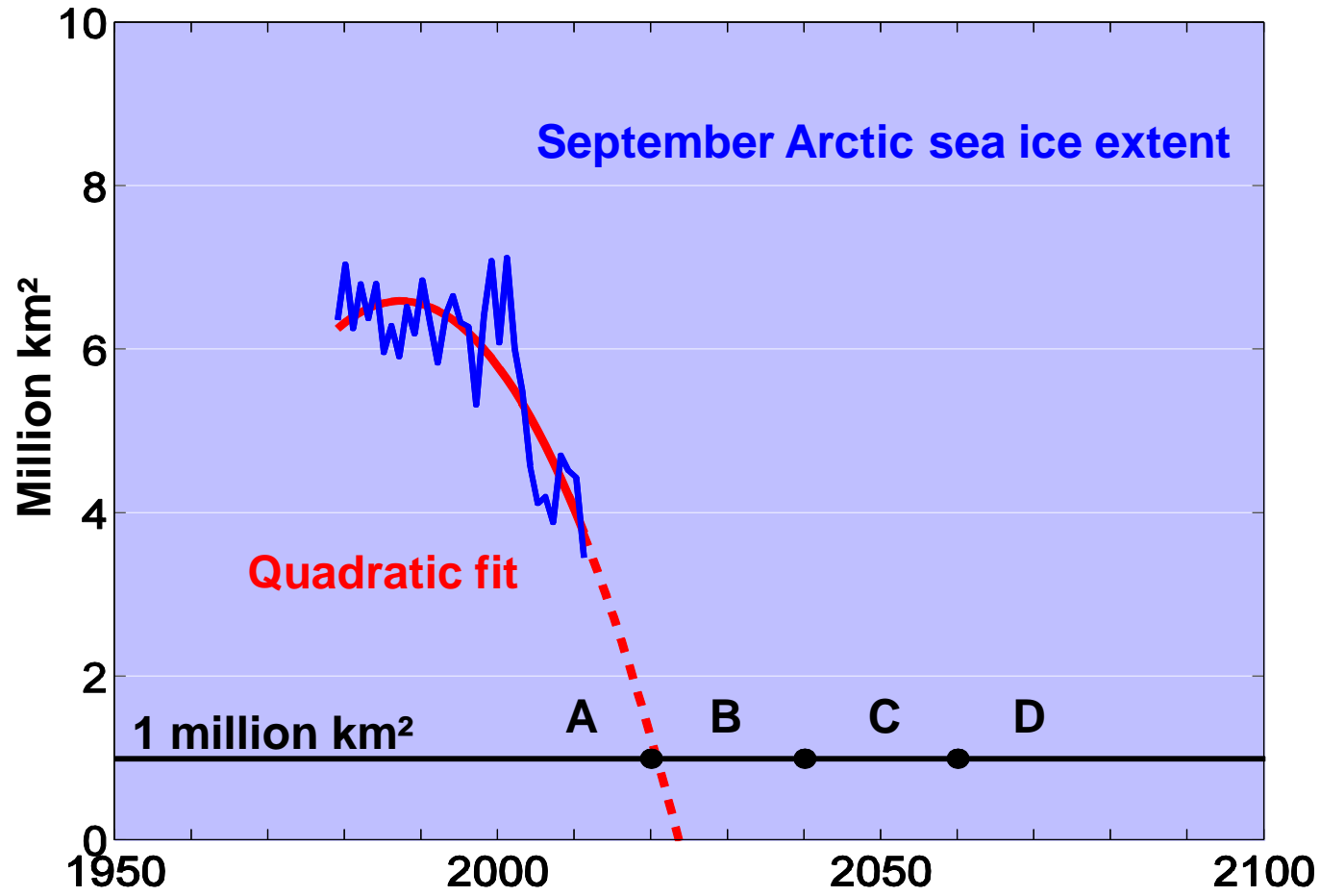
Challenge 2

Summer Arctic sea ice predictability



Challenge 2

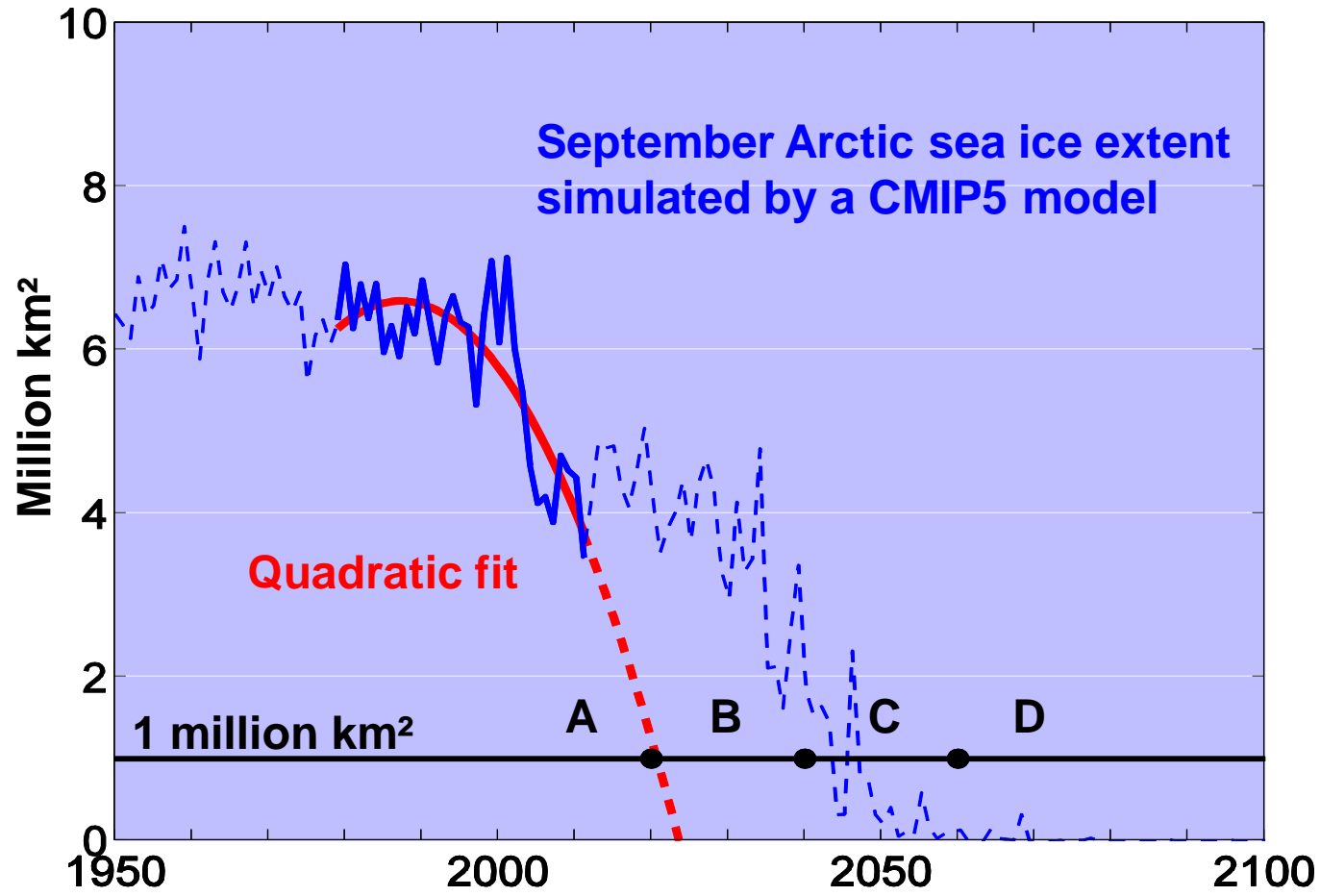
Summer Arctic sea ice predictability



Challenge 2

Summer Arctic sea ice predictability

Never rely on your first intuition for complex systems!



Data assimilation and ensemble Kalman filter

Reconstructing Antarctic sea ice changes

Is there an added value from sea ice initialization?

The ensemble Kalman filter is a multivariate data assimilation method

$$\begin{array}{l}
 \text{Analysis} \\
 \mathbf{x}^a
 \end{array}
 =
 \begin{array}{l}
 \text{Forecast} \\
 \text{(NEMO-LIM)} \\
 \mathbf{x}^f
 \end{array}
 +
 \begin{array}{l}
 \text{Kalman gain} \\
 \mathbf{K}
 \end{array}
 \cdot
 \begin{array}{l}
 \text{Observations} \\
 \text{Global, daily sea-} \\
 \text{ice concentrations} \\
 \text{[OSISAF - met.no]} \\
 \mathbf{d} - \mathbf{H} \mathbf{x}^f
 \end{array}$$

The diagrammatic representation below the equation uses colored rectangles to represent the variables:

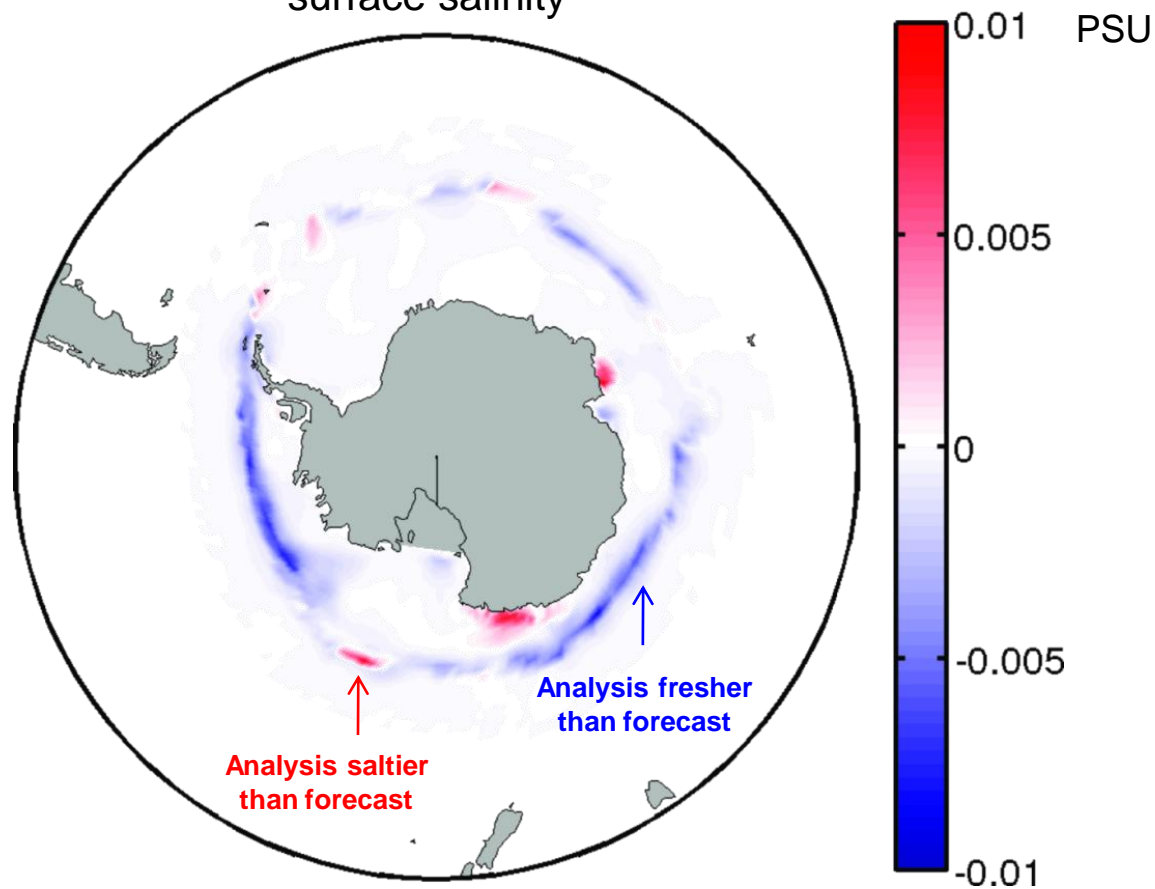
- An orange vertical bar represents the analysis vector \mathbf{x}^a .
- A blue vertical bar represents the forecast vector \mathbf{x}^f .
- A red vertical bar represents the Kalman gain matrix \mathbf{K} .
- A green vertical bar represents the observation vector \mathbf{d} .
- A purple horizontal bar represents the observation operator matrix \mathbf{H} .
- A blue vertical bar represents the forecast vector \mathbf{x}^f again, as part of the observation residual calculation.

 The equation is shown as:

$$\text{orange bar} = \text{blue bar} + \text{red bar} \cdot (\text{green bar} - \text{purple bar} \cdot \text{blue bar})$$

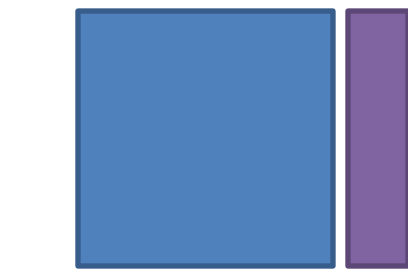
The ensemble Kalman filter is a multivariate data assimilation method

Example of an update in sea surface salinity

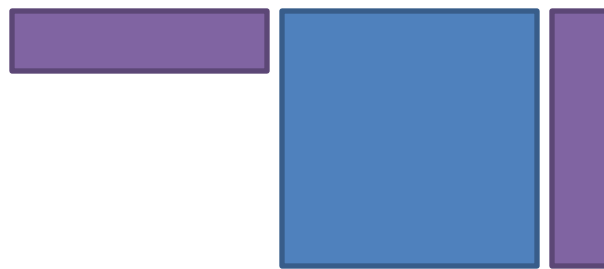


The ensemble Kalman filter relies on ensemble simulations

$$K = P H^T (H P H^T + R)^{-1}$$

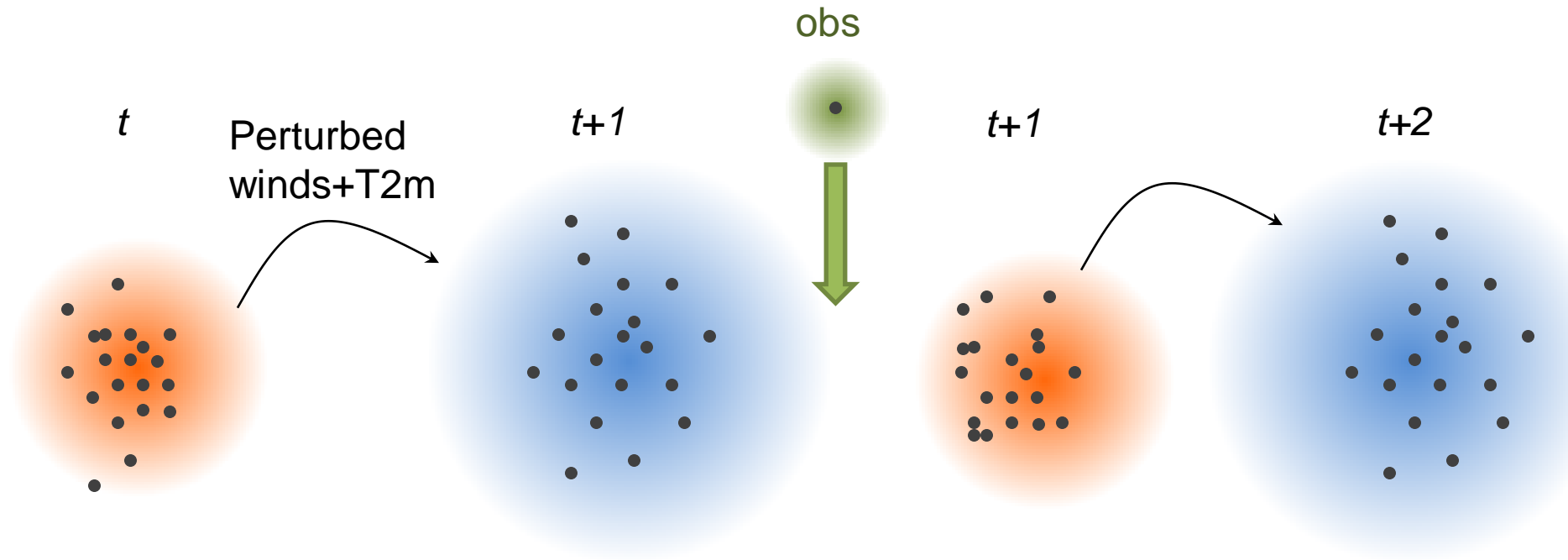


Model error
covariance matrix
(from 25 members)



Obs. error
covariance matrix
(diagonal for now)

The ensemble Kalman filter is a forecast-analysis method



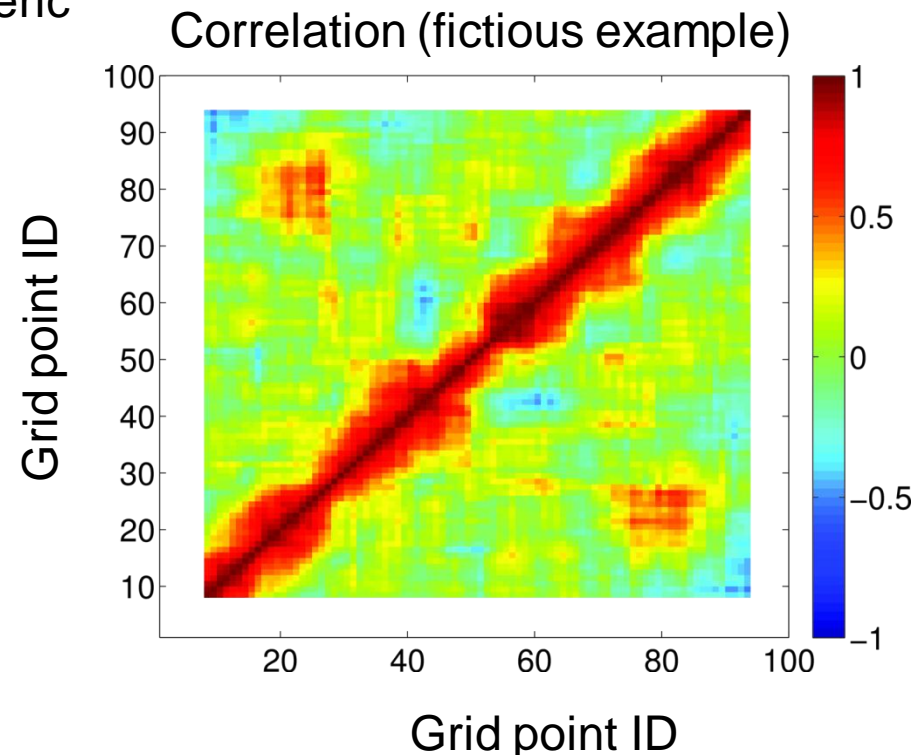
Ensemble spread, restartability and limitations

The distribution of ensemble members
should reflect the full model uncertainty

* 25 members with perturbed atmospheric
forcing (winds/2m-air temperature)

* Localization [Sakov and Bertino, 2010]

* Use of perturbed observations
[Burgers et al., 1998]



Ensemble spread, restartability and limitations

A « sanity check » for the model is necessary because gaussianity assumption is rarely fulfilled

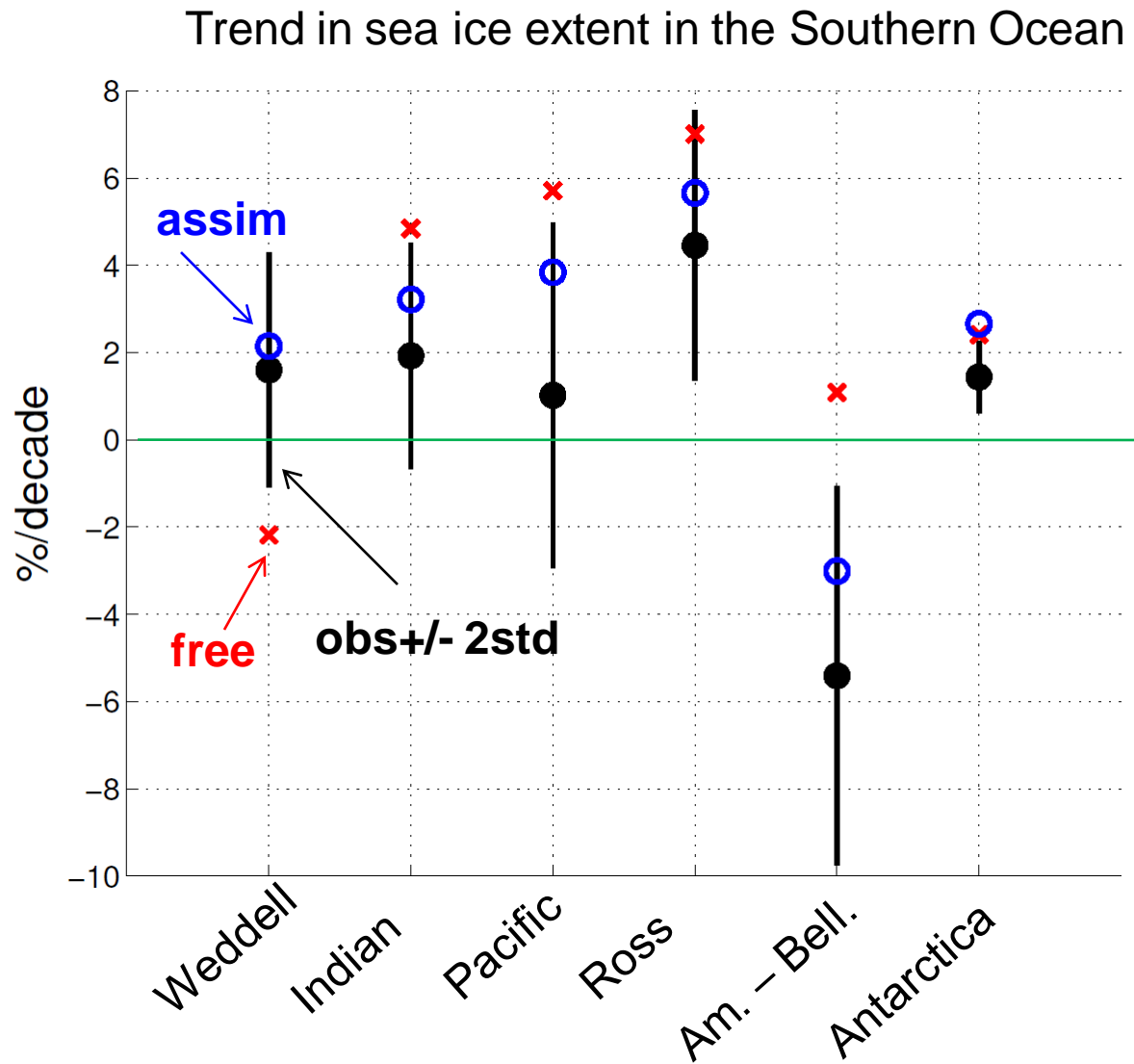
- > Reset negative ice concentrations/thickness to zero
- > Bound total ice concentration by 1
- > Ice thickness stays within category bounds

Data assimilation and ensemble Kalman filter

Reconstructing Antarctic sea ice changes

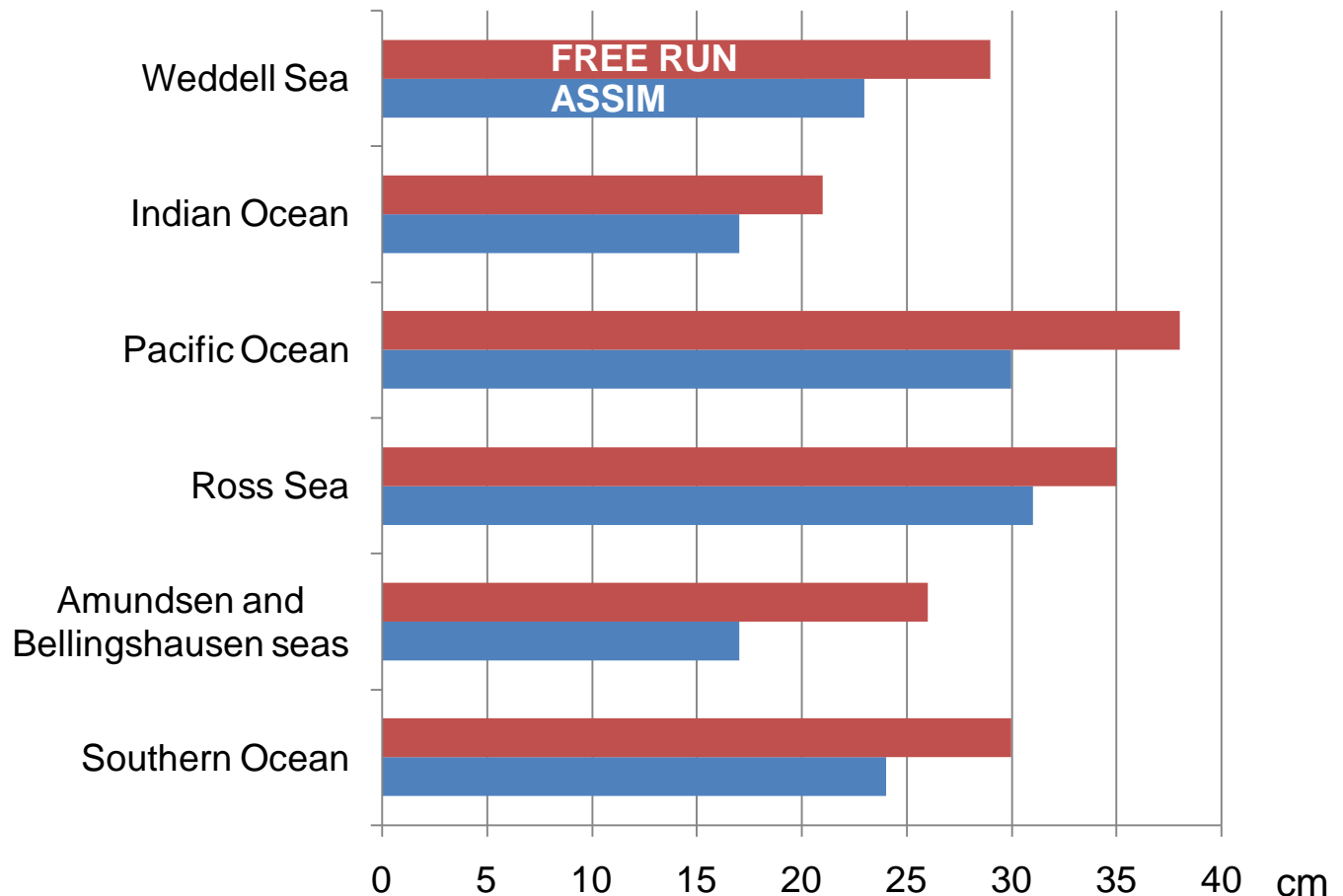
Is there an added value from sea ice initialization?

Improved sea ice extent variability



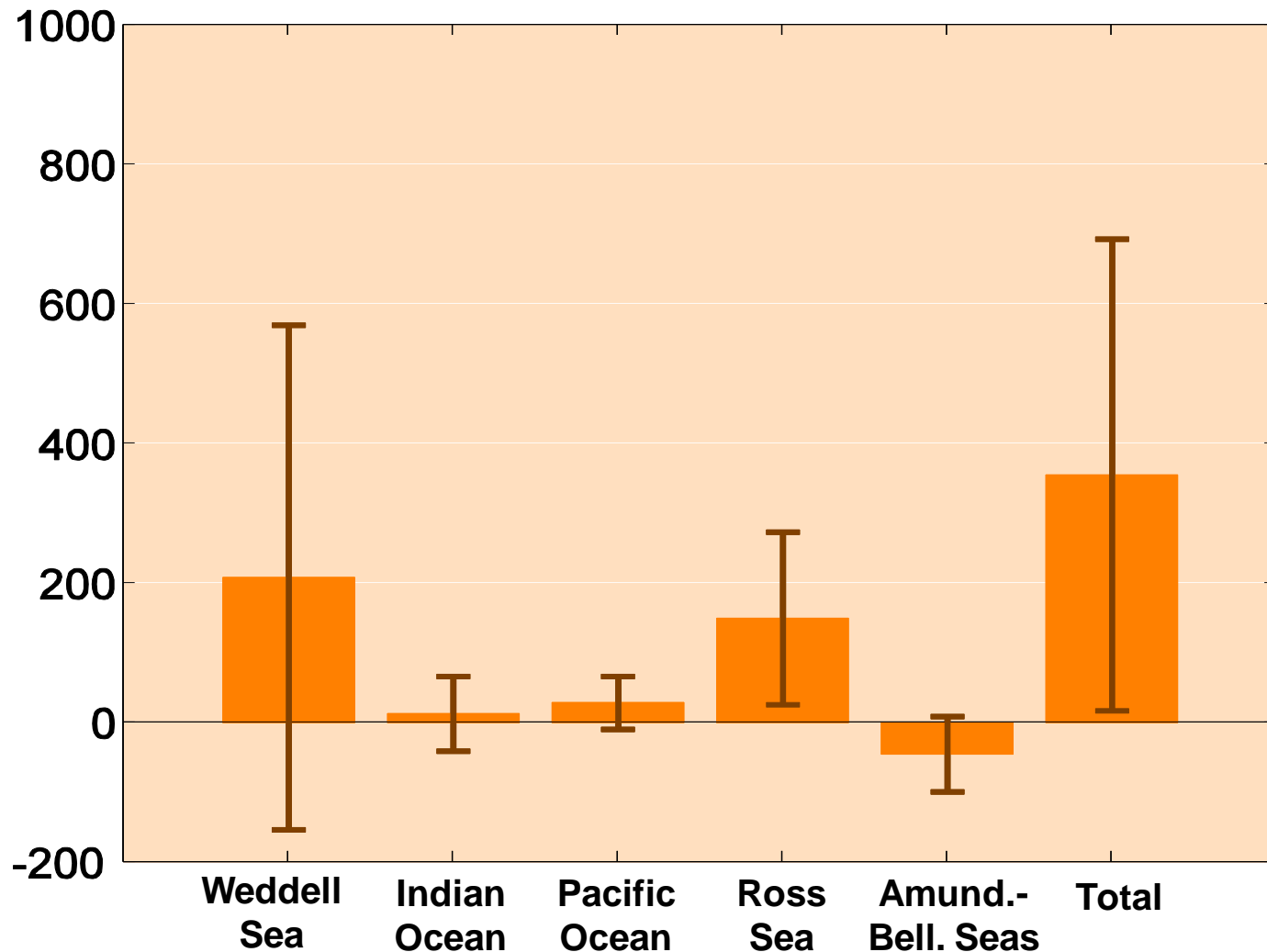
Improved simulated sea ice thickness

Mean **bias** in simulated thickness against ASPeCt data [Worby et al., 2008]



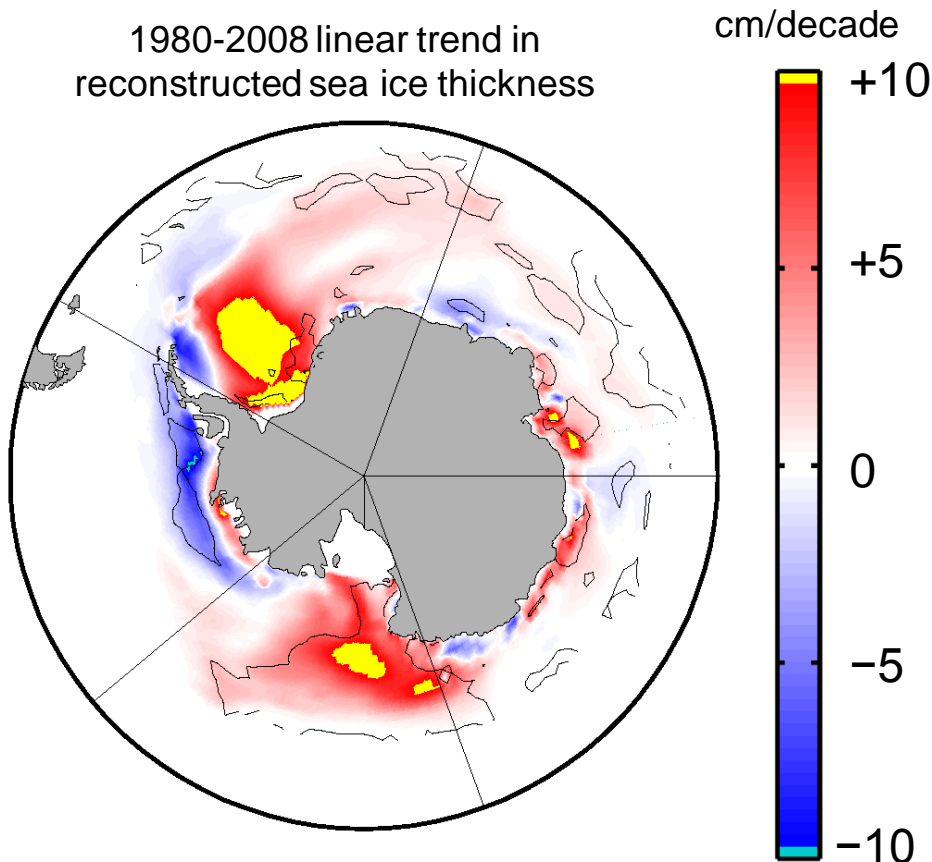
Weak, regionalized and noisy increase in Antarctic sea ice volume

1980-2008 trend in
sea ice volume
[km³/decade]



Mechanisms for Southern Ocean sea ice variability

1980-2008 linear trend in reconstructed sea ice thickness



The global increase in volume should be analyzed at the regional scale first

Regional signed responses are a result of regional dynamical and thermodynamical processes

- Changes in winds [Holland and Kwok, 2012]
- Changes in hydrological cycle [Zhang, 2007; Bintanja et al., 2013]
- Unforced variability [Polvani & Smith, in rev.]
- ...

Data assimilation and Ensemble Kalman filter

Reconstructing Antarctic sea ice changes

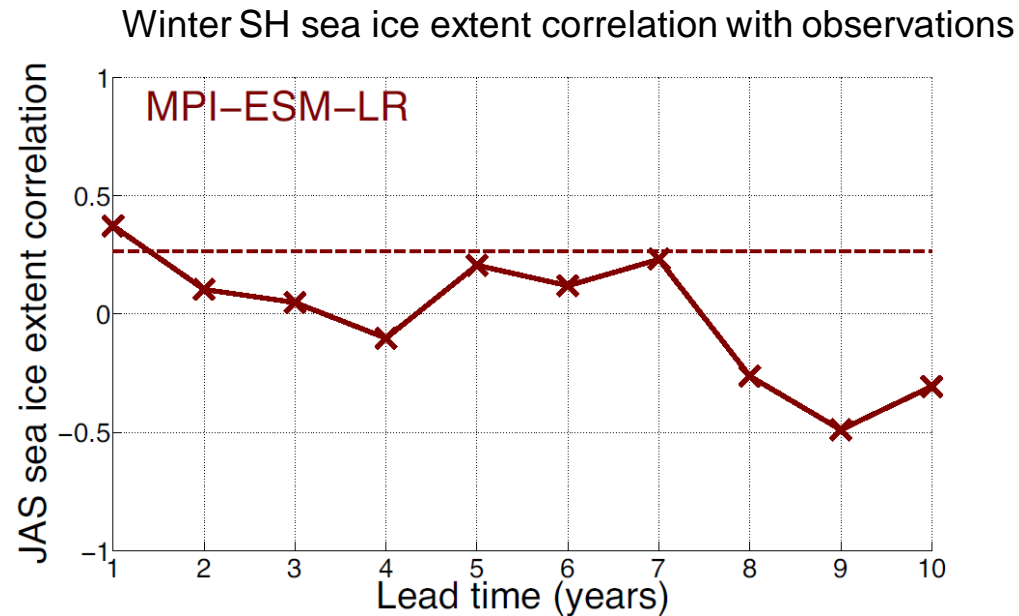
Is there an added value from sea ice initialization?

The predictive skill from a model may depend on

- > The metrics used to measure it
- > The model used
- > The data assimilation method used

Antarctic ocean/sea ice initialization efforts

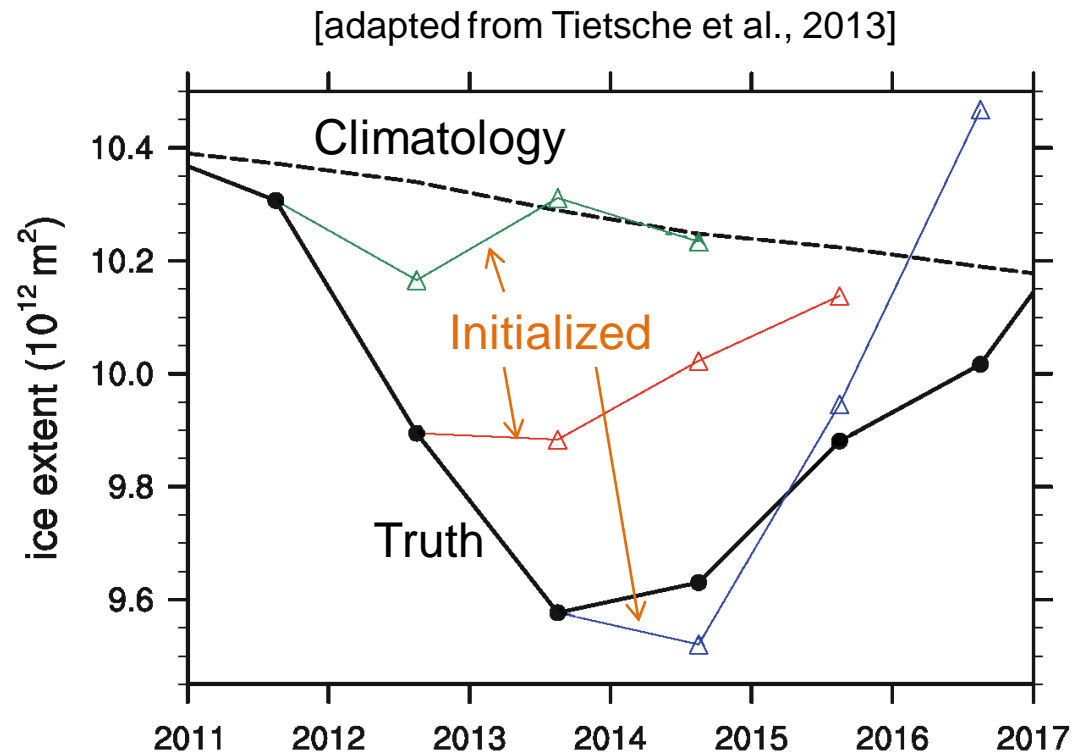
Initialization of 3-D ocean does not significantly improve sea ice predictability



[Zunz et al., 2013]

Arctic sea ice initialization efforts

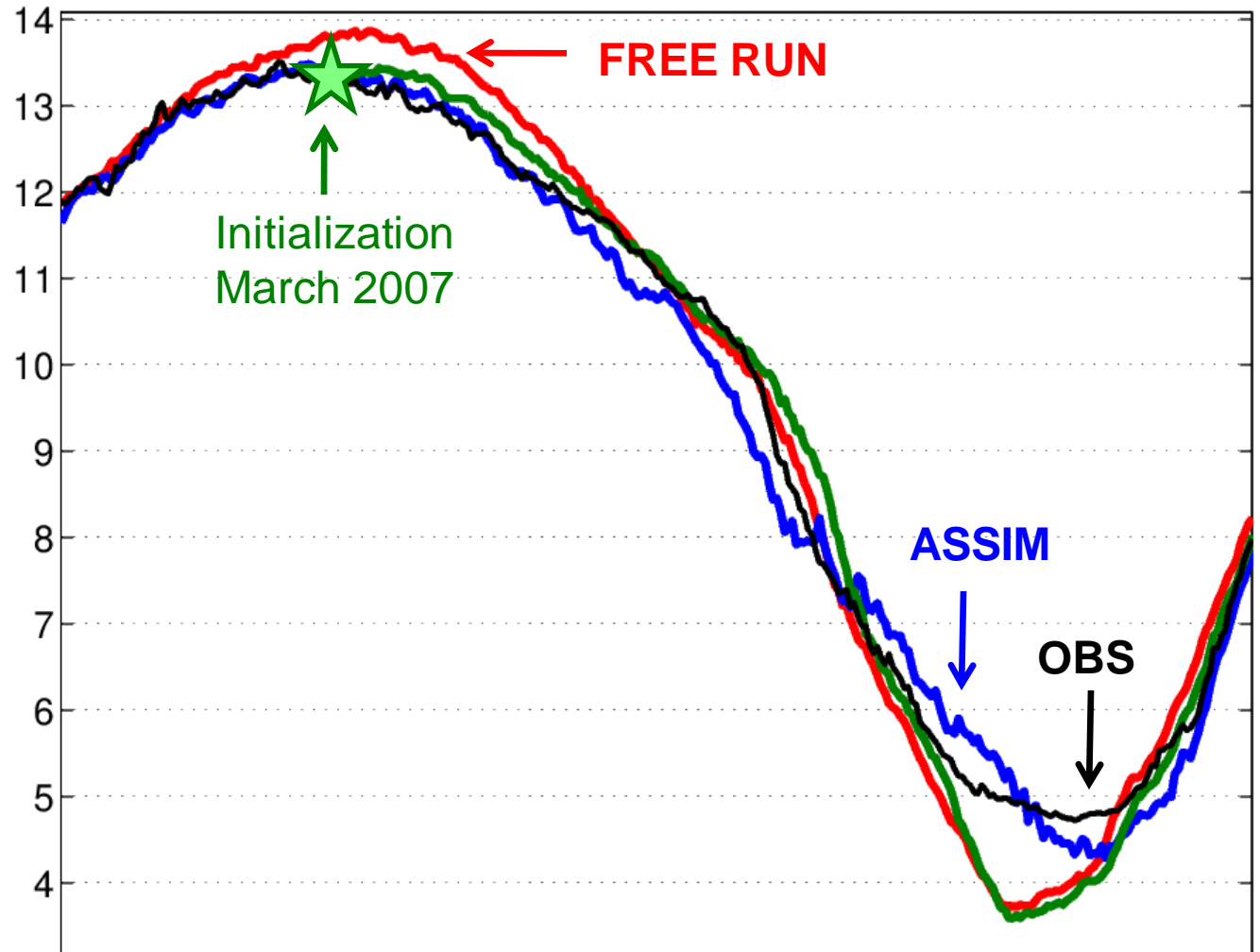
The onset, timing and amplitude of large sea ice anomalies are hardly predictable



Arctic sea ice initialization efforts

Year 2007

Arctic sea ice
extent [million km²]



Conclusions

Arctic & Antarctic sea ice data assimilation

Useful for sea ice state reconstruction

Marginal improvements for sea ice
predictability

Prospects for joint state/parameter
estimation

Thank you

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