

ERA-CLIM2 Project

Mercator Ocean Contributions to WP2.2

Task 2.2 : Development of assimilation techniques



**Mercator
Ocean**
Ocean Forecasters

C.-E. Testut,
C. Bricaud, J. Chanut,
G. Garric, G. Ruggiero

...

WP2 task2.2 : Development of assimilation techniques for improved use of surface observations

MO Sub task : Assimilation of Sea Ice Concentration

Objective: to improve coupled ocean/sea-ice data assimilation.
Developing and testing a scheme that transforms sea-ice concentration to a Gaussian variable during the assimilation process.

Deliverable D2.2 (t0+27+12):

Results from a study
Documented code and library.

1 - Production of reference reanalysis without the use of Gaussian transformation in the sea ice analysis

→ Development of a multivariate sea ice analysis vs univariate sea ice analysis (used in operational)

⇒ **Well posed to demonstrate the interest of the Gaussian transformation due to the presence of wrong extrapolation**

→ Use the Arctic-Northern Atlantic Configuration at $1/4^\circ$ (75lev, CREG4/NEMO3.6/LIM3) coupled with the Mercator Assimilation System (SAM2) and the multivariate/univariate sea ice analysis.

⇒ **a low-cost and recent model configuration centred on the Arctic Sea is more efficient for this study...**

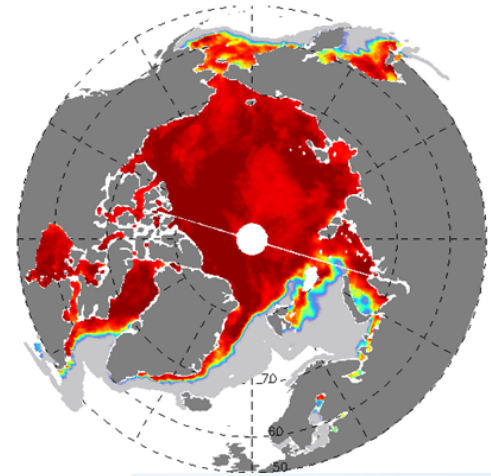
We have realized various multi-years reanalysis using different tuning for the state vector and we have identified a first robust set up for a multivariate sea ice analysis.

Multivariate Sea Ice Reanalysis

Main characteristics



Sea Ice Concentration (OSI-SAF)



Model

- Nemo 3.6, LIM3/Multi-categories(1:15)
- CREG 1/4, 75 levels

Assimilation

- Analysis based on a 2D local multivariate SEEK/LETKF filter
- Weakly-coupled DA system using 2 separate analyses :
- Ocean Analysis (SLA, InSitu Data from CORA3.2, SST) , IAU on (h,T,S,U,V)
- Sea Ice Analysis
 - SIC Error: 1% open ocean, linear from 25% to 5% for SIC values between 0.01 and 1
 - Forecast error covariances are built from a prior ensemble of Sea Ice Model anomalies
- Unidata/Multivariate Sea Ice analysis
 - CREG4 reanalysis using multivariate state vector [SIC, SICONCAT(1:15), SIVOLUCAT(1:15)]

Multivariate Sea Ice Reanalysis

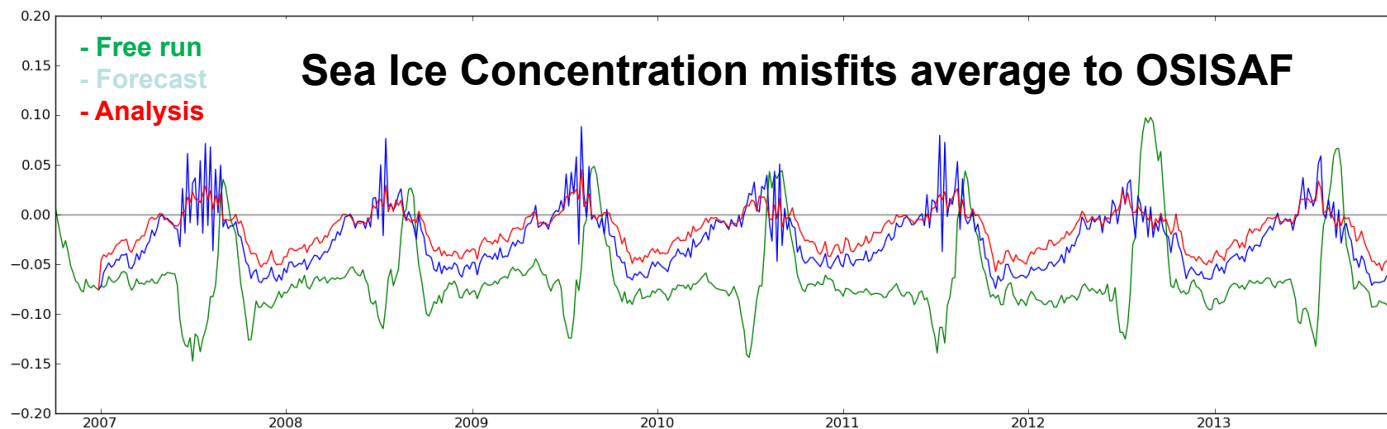
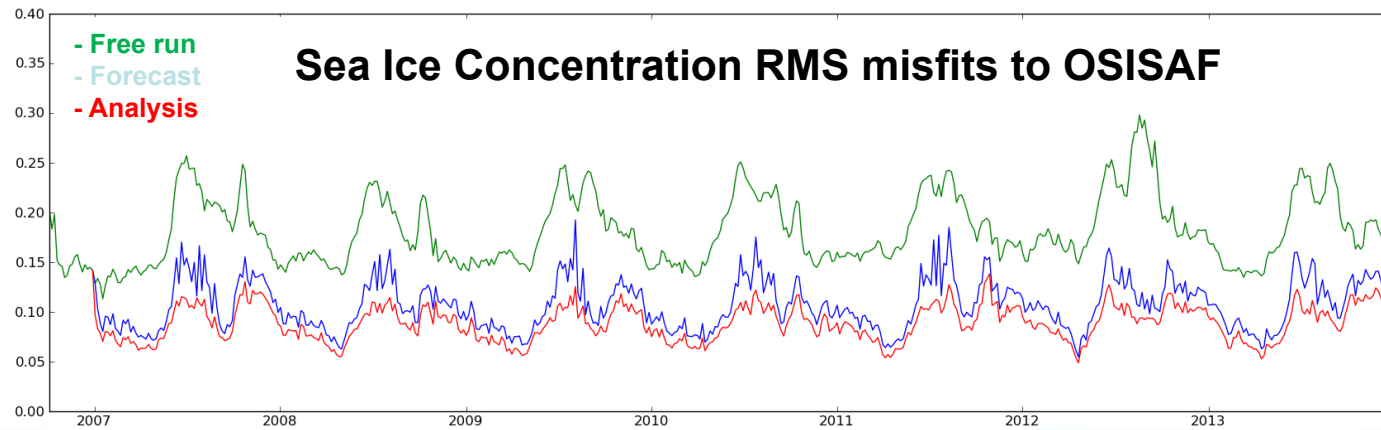
CREG4 reanalysis assimilating OSI-SAF SIC Observations



CREG4 Reanalysis starting at 20070102

Model update from analysis : [siconc,siconcat,sivolucat]

(Δ siconcat(1:15), Δ sivolucat(1:15)) estimated from analysis based on OSI-SAF SI and null innovations on SICONCAT(15) and SIVOLUCAT(15)



Multivariate Sea Ice Reanalysis

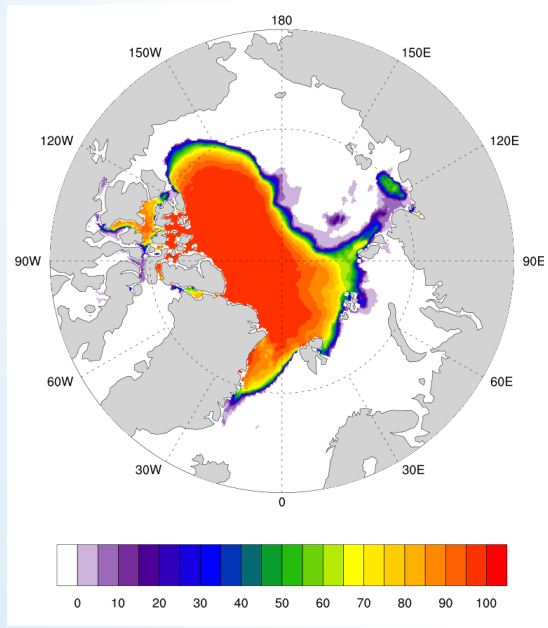
CREG4 reanalysis assimilating OSI-SAF SIC Observations



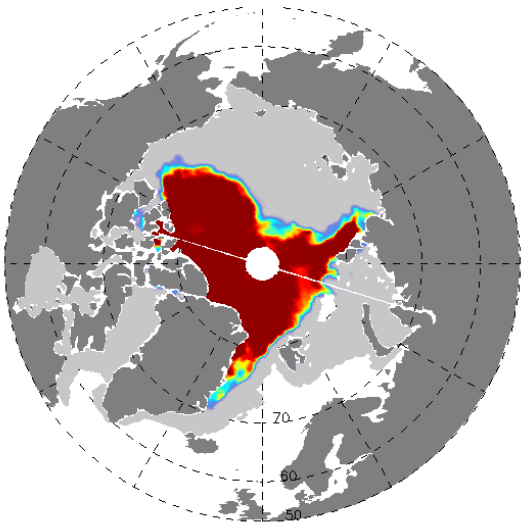
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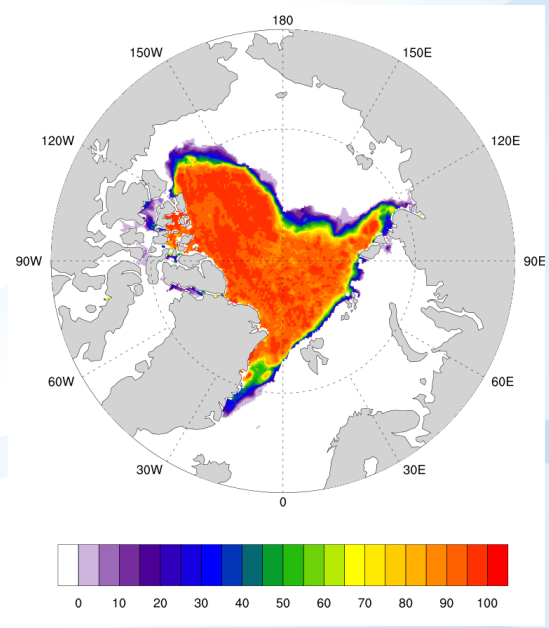
SIC(%) 20070915



CREG4 free run



OSI-SAF



CREG4 reanalysis

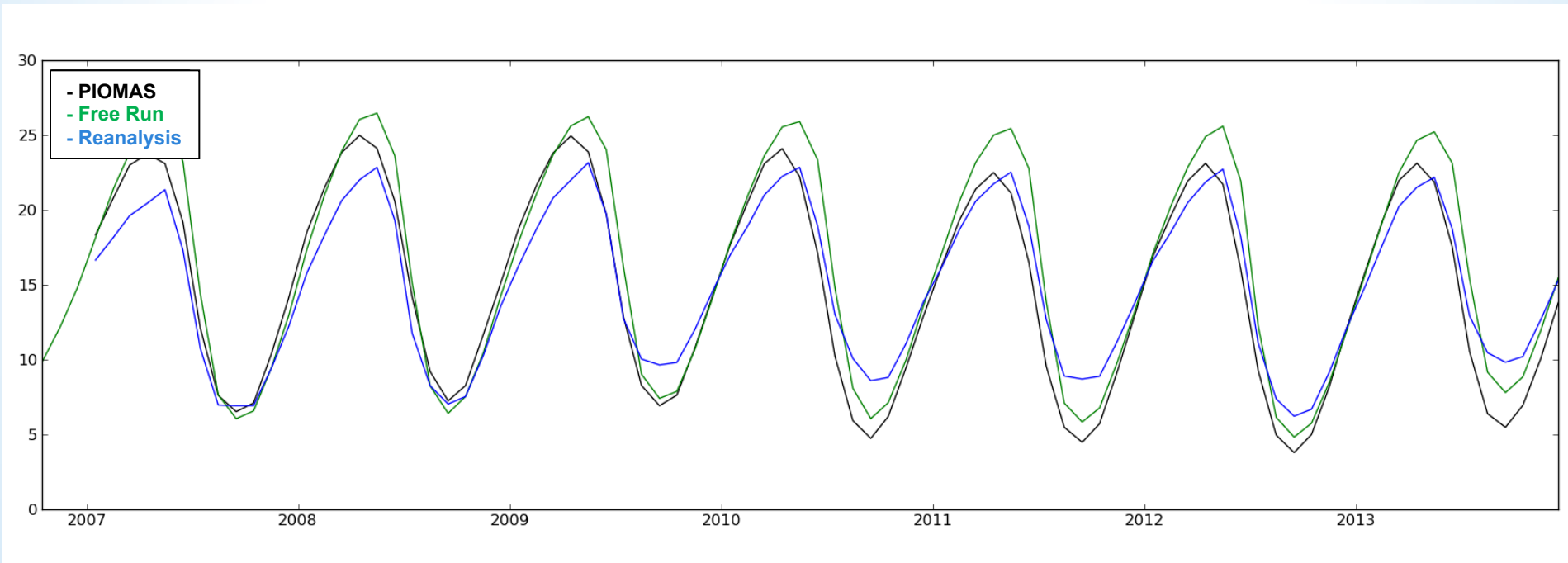
Multivariate Sea Ice Reanalysis

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Sea Ice Volume (Arctic)

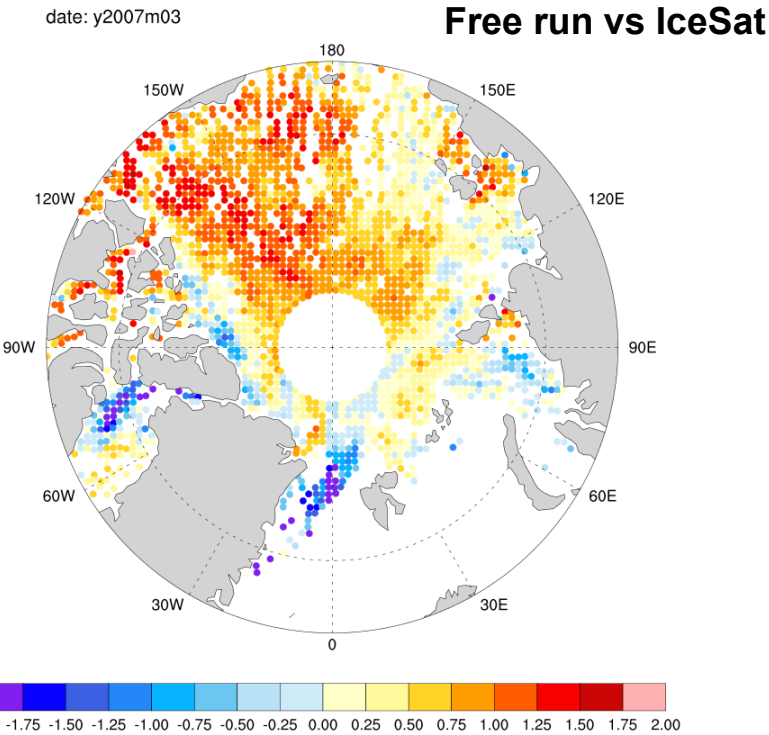
Multivariate Sea Ice Reanalysis

CREG4 reanalysis assimilating OSI-SAF SIC Observations

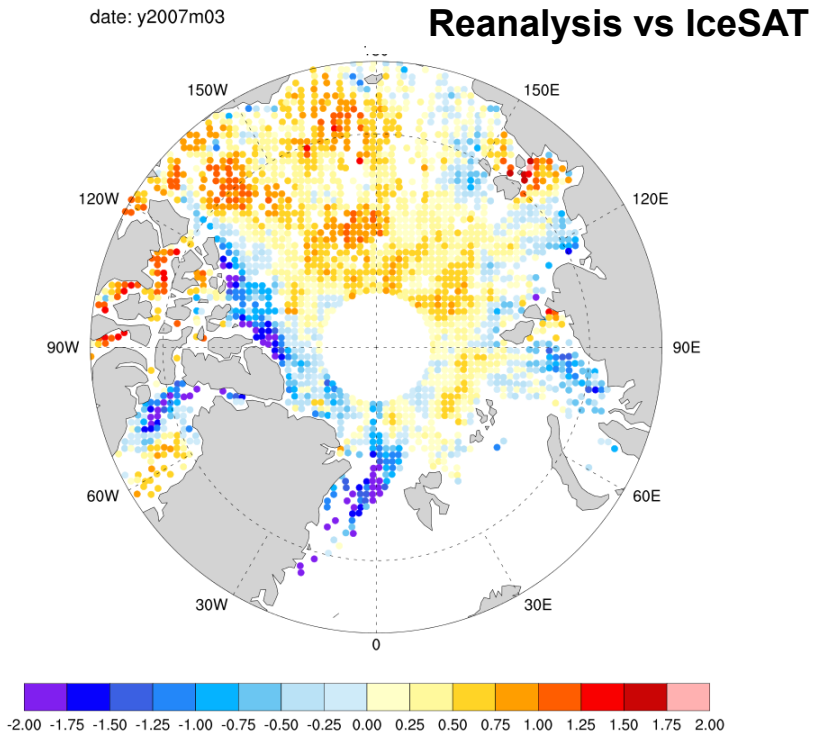


CREG4 reanalysis starting at 20070102
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Sea ice thickness difference at y2007m03



Sea ice thickness difference at y2007m03



1 - Production of reference reanalysis without the use of Gaussian transformation in the sea ice analysis

- Development of a multivariate sea ice analysis vs univariate sea ice analysis (used in operational)
 - ⇒ **Well posed to demonstrate the interest of the Gaussian transformation due to the presence of wrong extrapolation**
- Use the Arctic-Northern Atlantic Configuration at $1/4^\circ$ (75lev, CREG4/NEMO3.6/LIM3) coupled with the Mercator Assimilation System (SAM2) and the multivariate/univariate sea ice analysis.
 - ⇒ **a low-cost and recent model configuration centred on the Arctic Sea is more efficient for this study...**

We have realized various multi-years reanalysis using different tuning for the state vector and we have identified a first robust set up for a multivariate sea ice analysis.

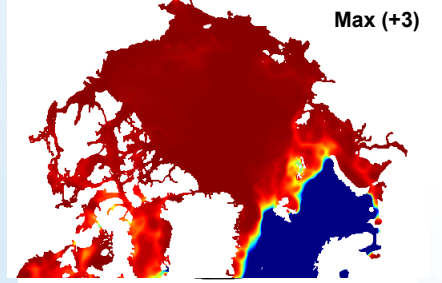
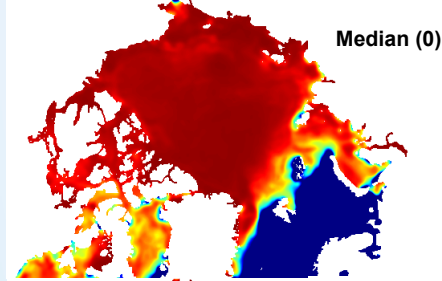
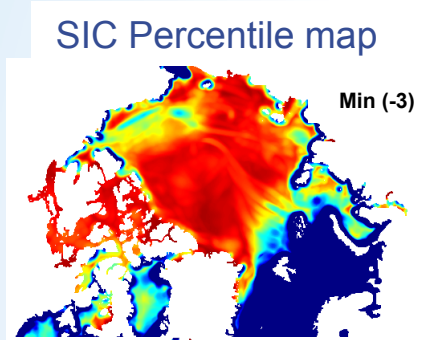
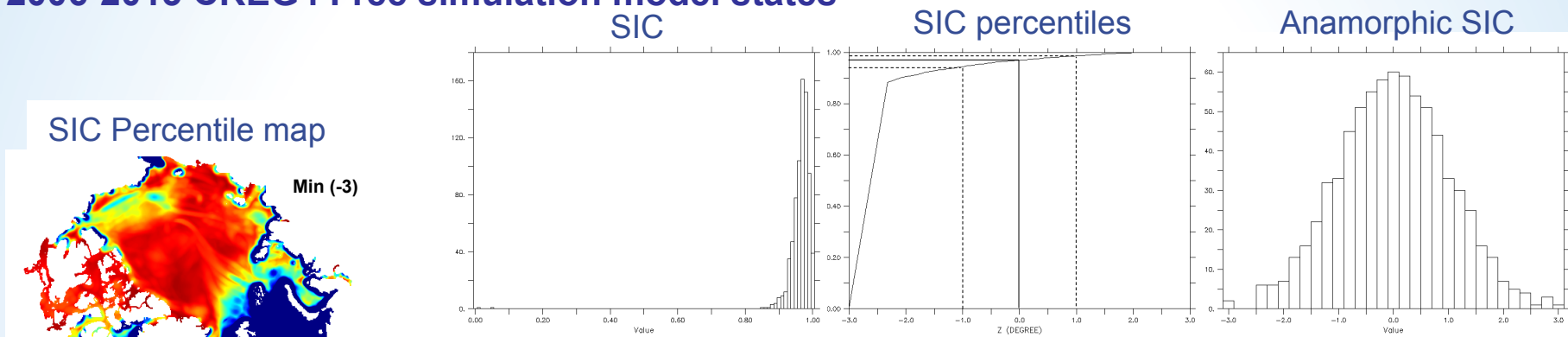
2 - Investigate the role of the Gaussian anamorphic transformation on the sea-ice data assimilation problem

- Idealized exercises with ensemble background error in order to evaluate the efficiency of sea-ice analysis
- realistic test case with the CREG4/NEMO3.6/LIM3 model using 256 members
 - ⇒ **1D multivariate (SIC, VOL) sea ice analyses assimilating SIC only provided by a nature run**

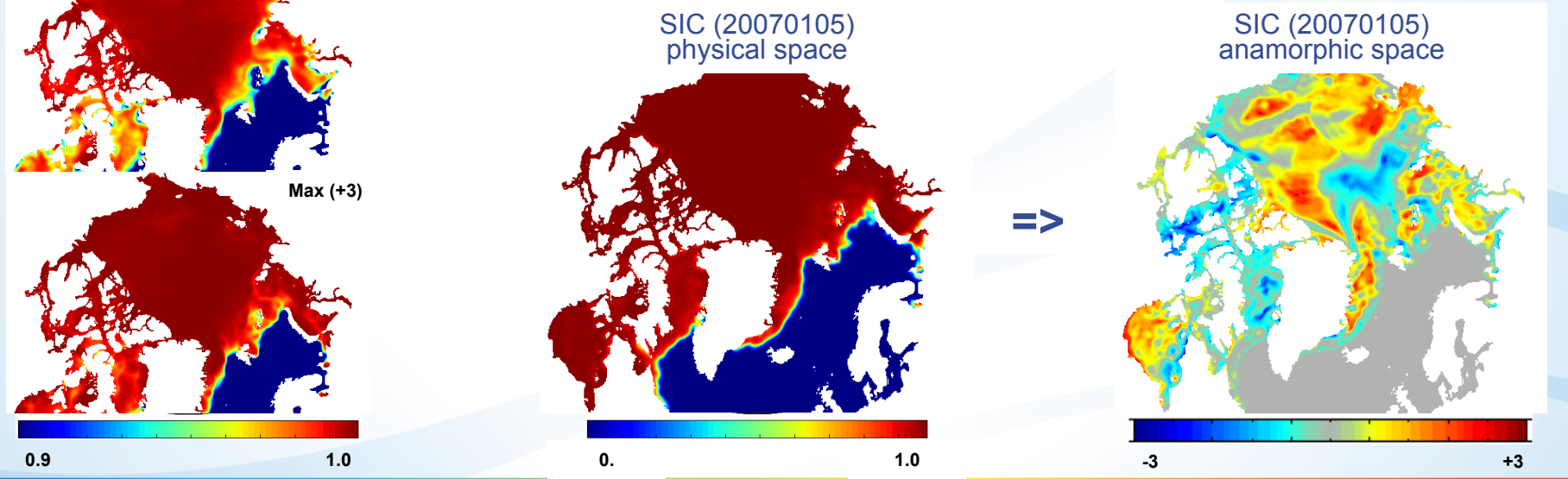
The Gaussian Anamorphosis approach

Anamorphic Transformation based on

2006-2013 CREG4 Free simulation model states



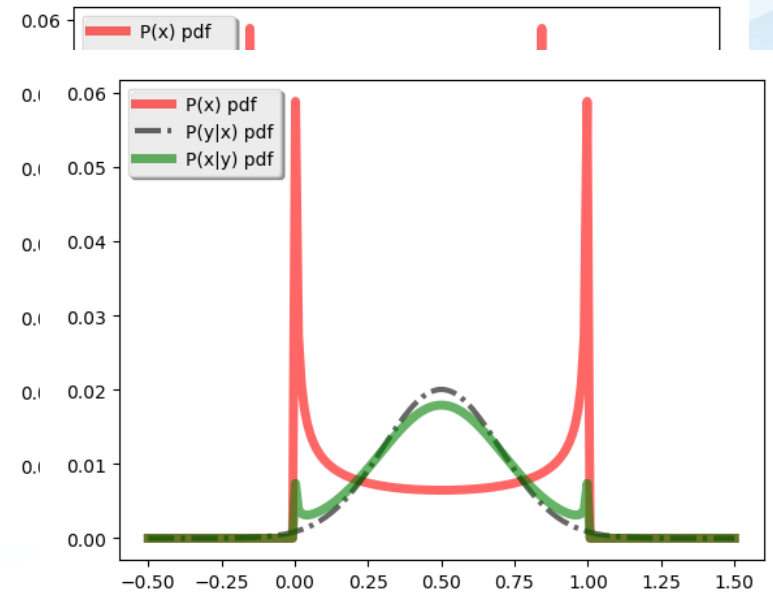
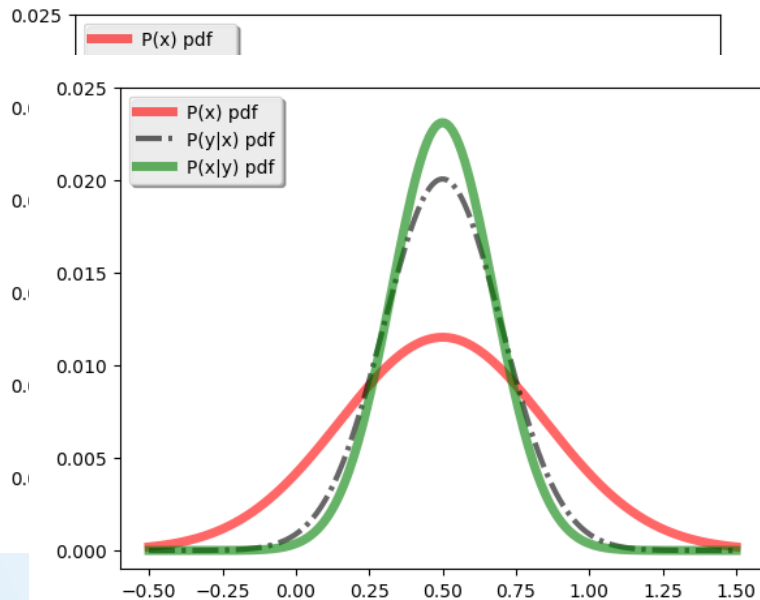
We build local (space and time) transformation for each variable of the state vector (SIC, SIVOLU, SICONCAT(1:15), SIVOLUCAT((1:15)))



Bayesian Data Assimilation aims to find the probability of the system state x given observations y of the system using the Baye's theorem:

$$p(x|y) = p(y|x) \cdot p(x)$$

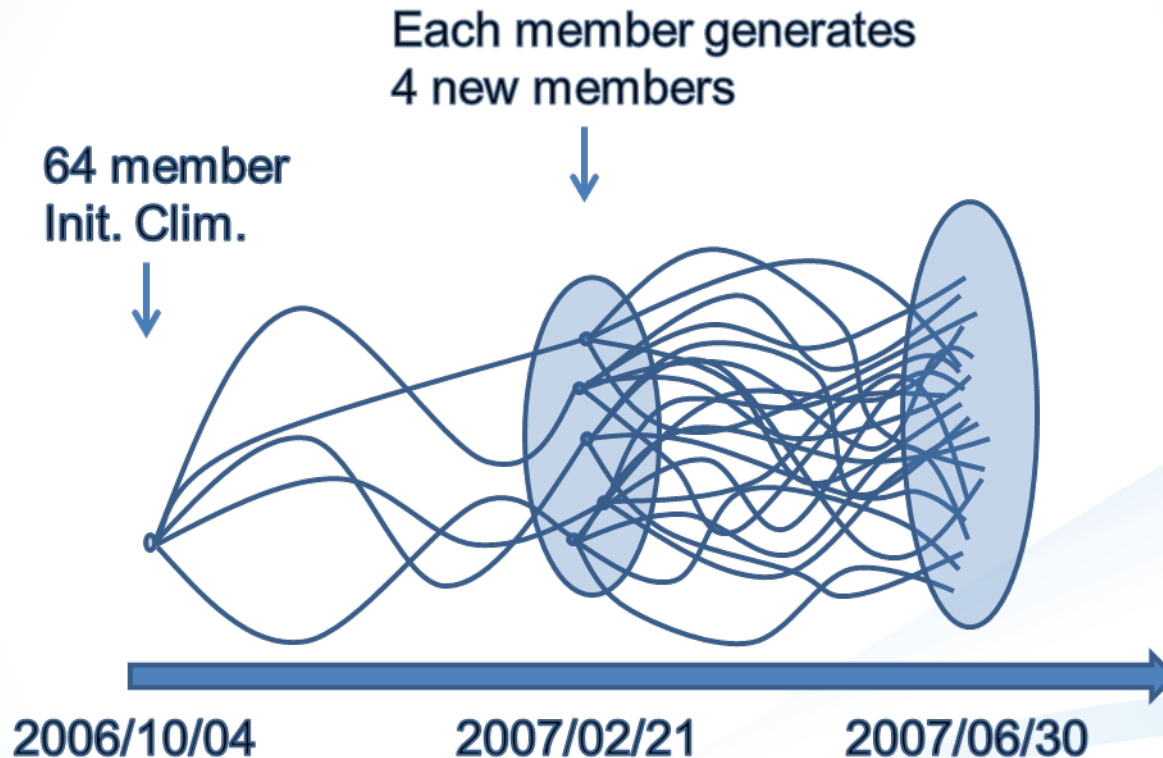
$y|x \sim N(0.5, 0.04)$ and $E[x] = 0.5$ and $\text{Var}[x] = 0.12$



Build idealized exercises comparing EnKF solution to the Bayesian solution

Realistic test case with CREG4/NEMO3.6/LIM3

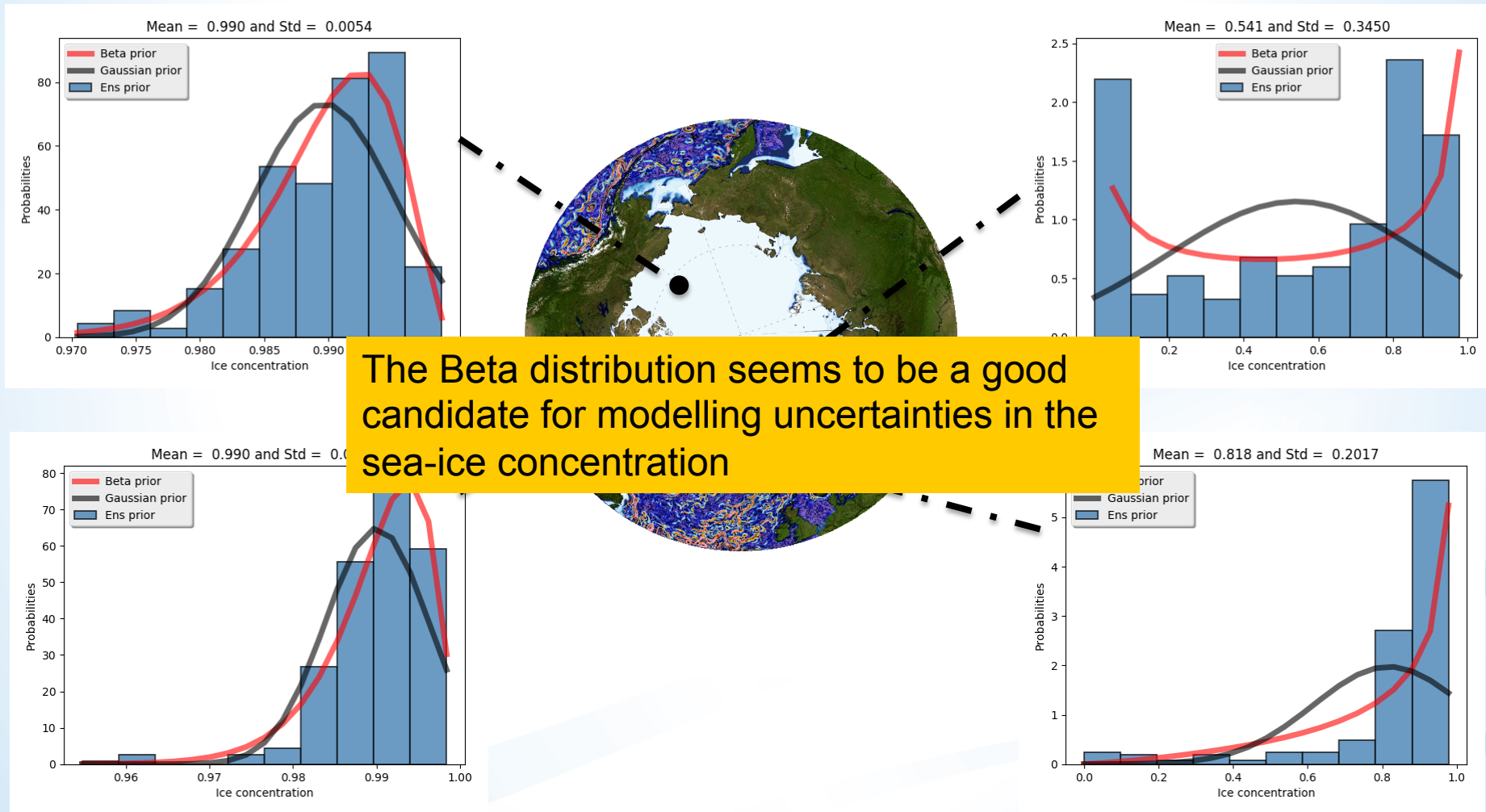
Sea-ice concentration prior distribution



A 256-members CREG4/NEMO3.6/LIM3 has been performed using perturbed rheology parameters and perturbed atmospheric fields and validated against distributions obtained using satellite observations.

256 members test case with CREG4/NEMO3.6/LIM3

Sea-ice concentration prior distribution

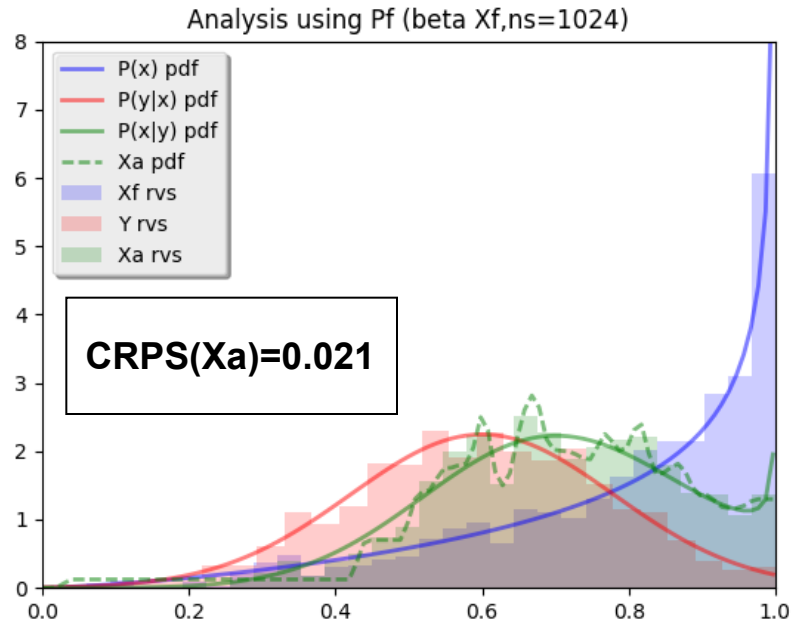


Idealized exercises

Data assimilation with a Beta distributed Prior

- How does the EnKF solution compares to the Bayesian solution?
- Can Gaussian anamorphosis help to partially restore bayesianity?

$$CRPS = \sum_{i=1}^{nbins} (cdf^B - cdf^F)^2$$



Idealized exercises

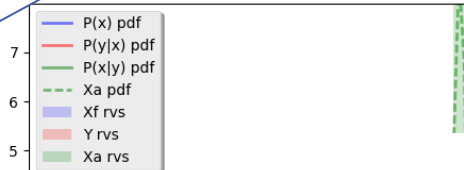
Data assimilation with a Beta distributed Prior

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ETkf

Enkf

Analysis using ETkf (beta Xf, ns=1024)



CRPS(Xa)=0.451

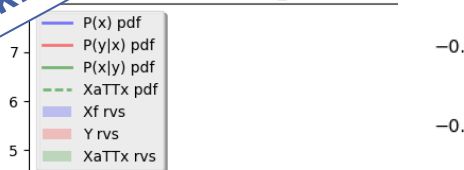
Analysis using Enkf (beta Xf, ns=1024)



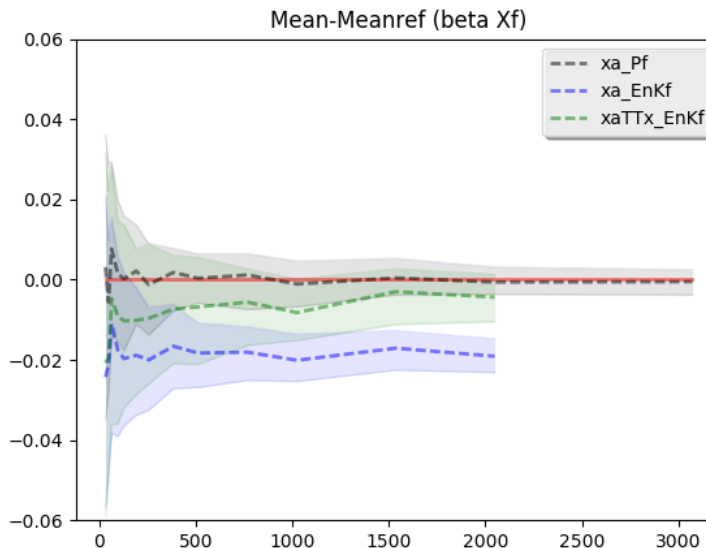
CRPS(Xa)=0.282

Enkf Tx

Analysis using Enkf_Tx (beta Xf, ns=1024)



CRPS(Xa)=0.072



Analysis using Pf (beta Xf, ns=1024)

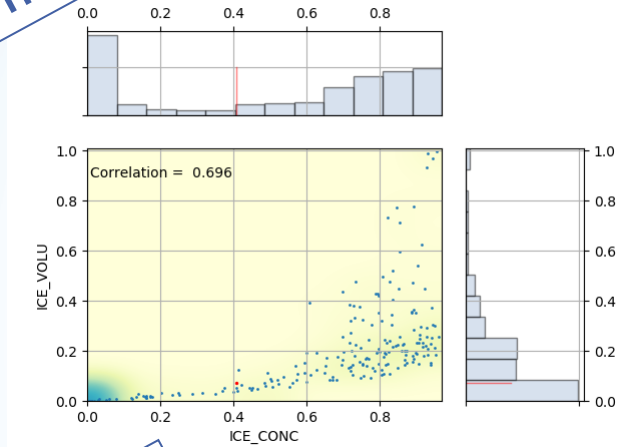


CRPS(Xa)=0.021

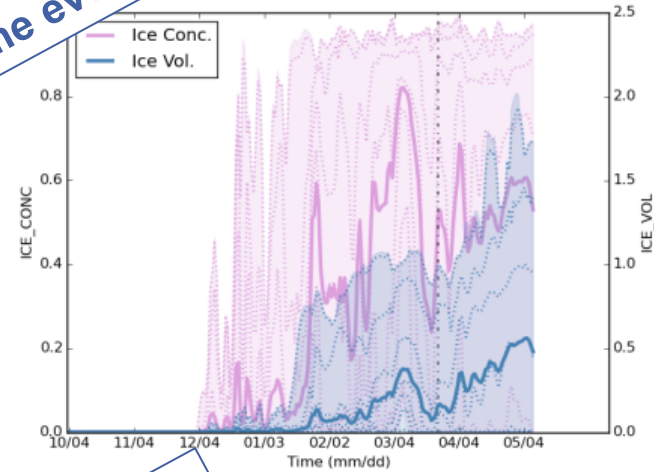
Realistic test case CREG4/NEMO3.6/LIM3

Concentration and Volume (256 members)

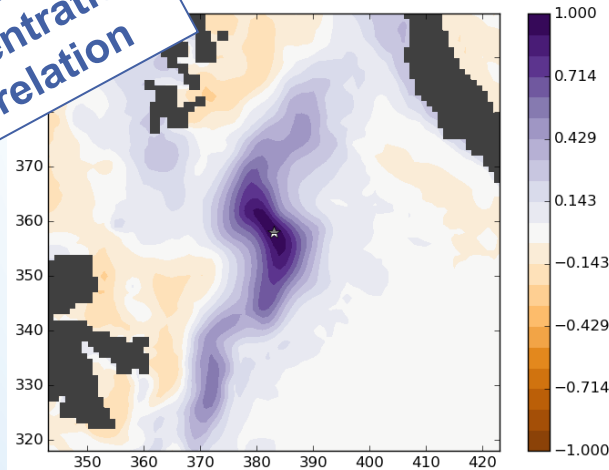
Prior



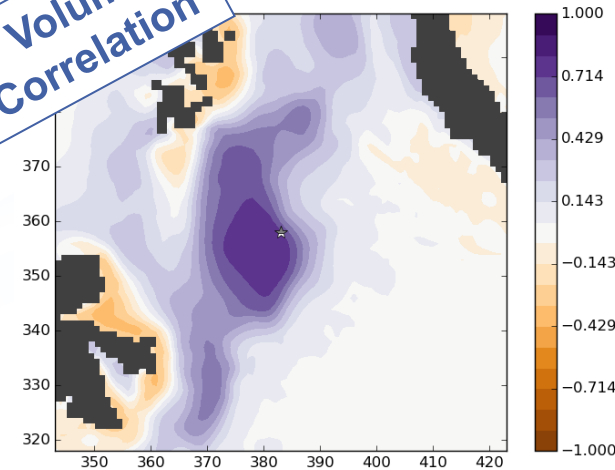
Time evolution



Concentration Correlation



Volume Correlation

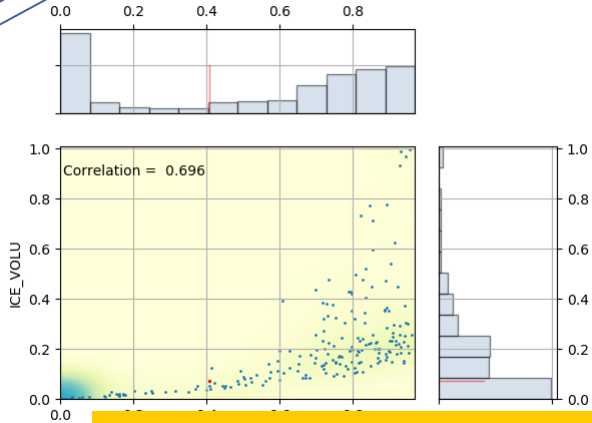


Realistic test case CREG4/NEMO3.6/LIM3

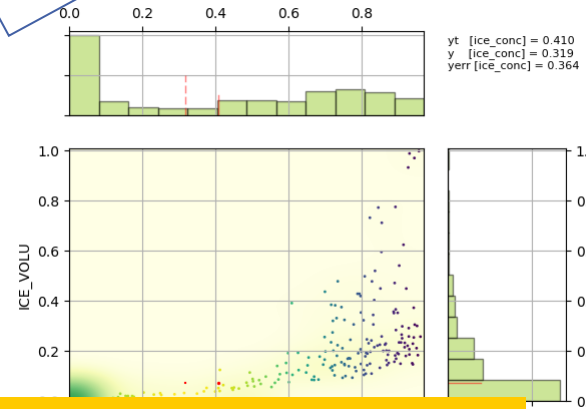
Bi-variate Concentration and Volume (256 members)

1D multivariate sea ice analysis assimilating SIC only provided by a nature run

Prior

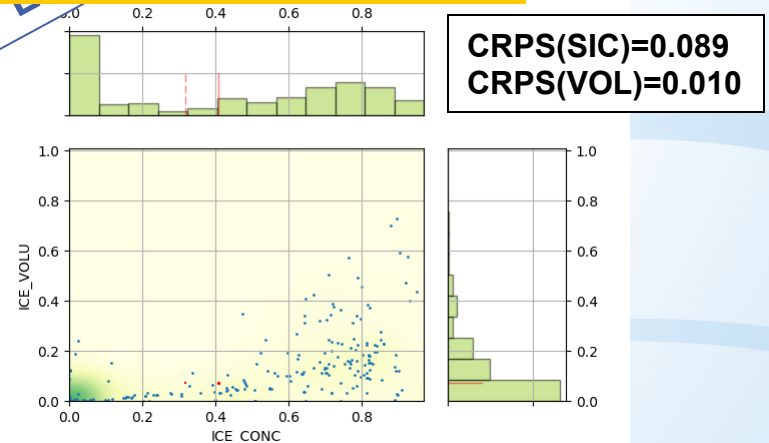
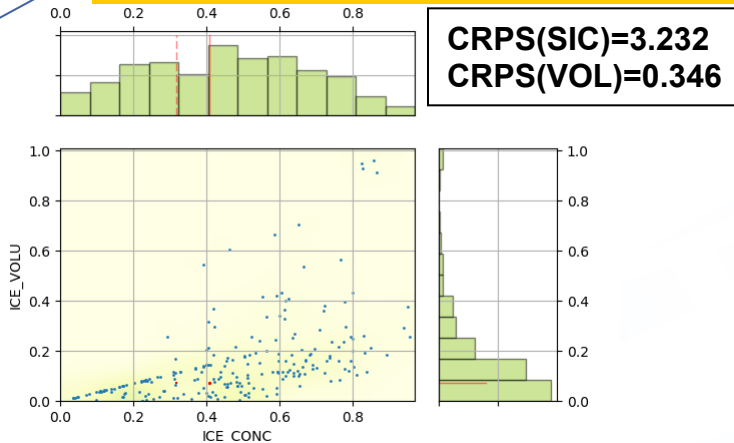


Pf



Gaussian anamorphosis improves the posterior distribution in particular in the sea ice extent

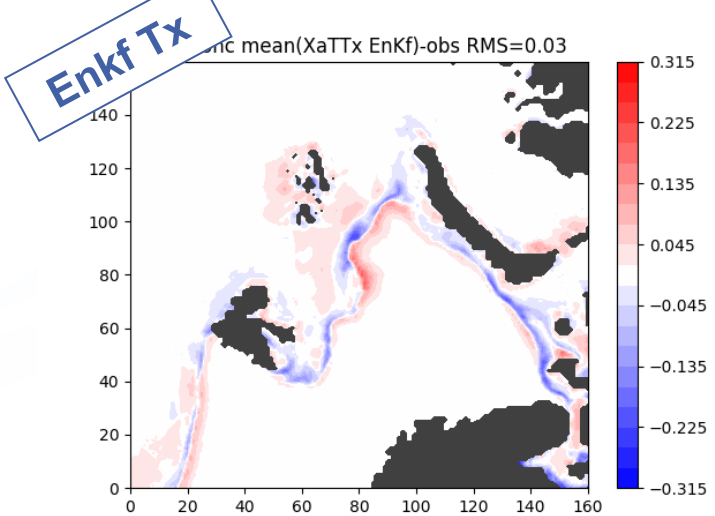
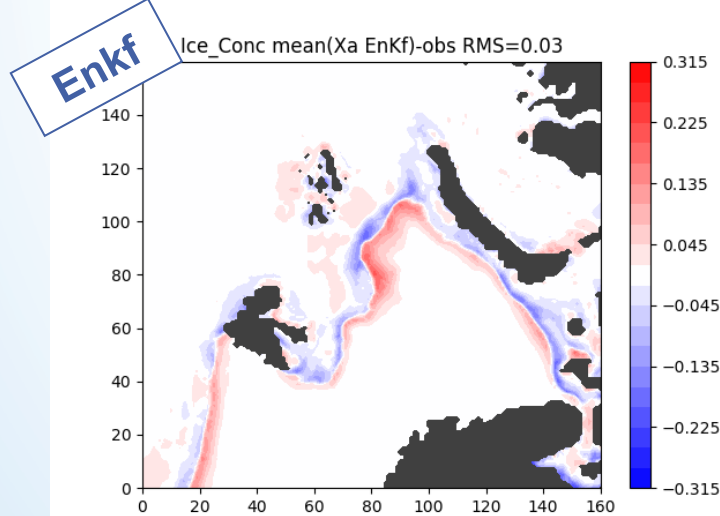
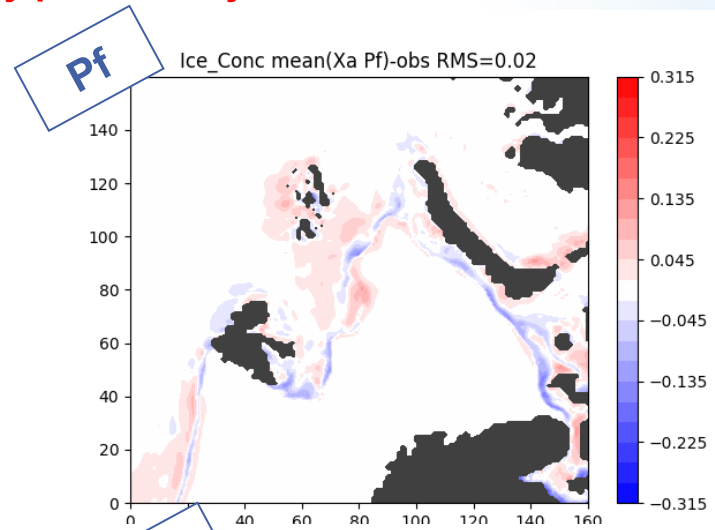
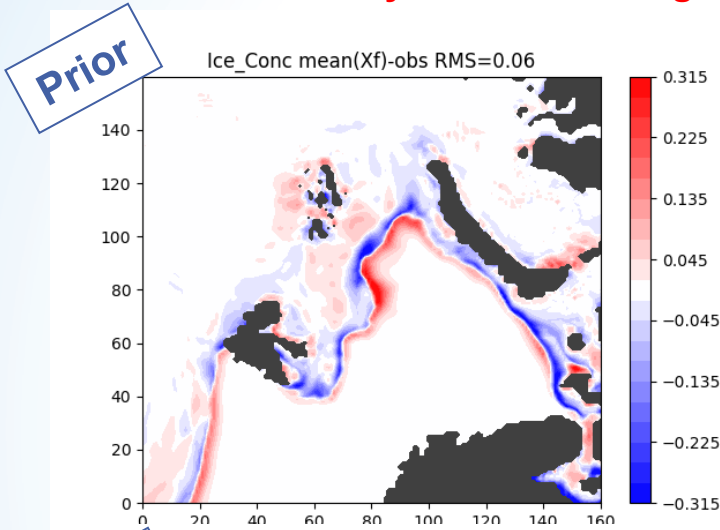
Enkf



Realistic test case CREG4/NEMO3.6/LIM3

Misfit of sample mean to nature run

1D multivariate sea ice analyses assimilating SIC only provided by a nature run



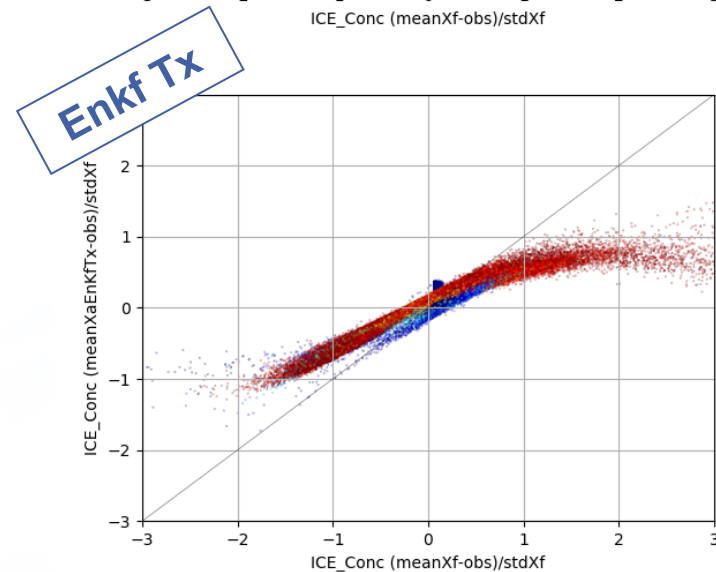
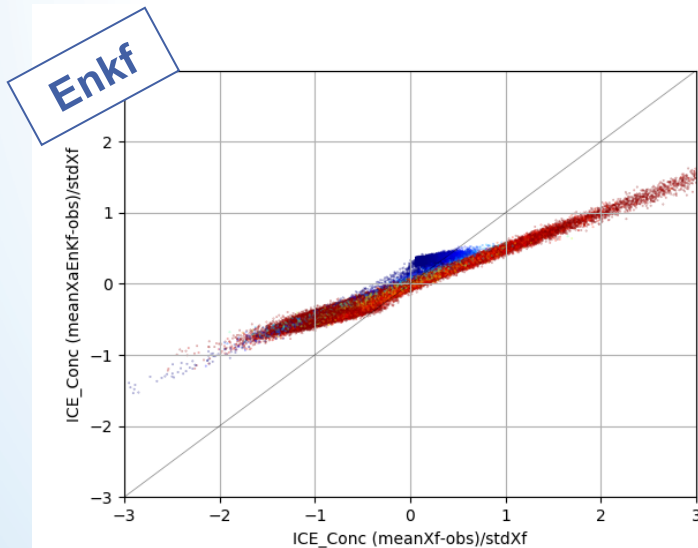
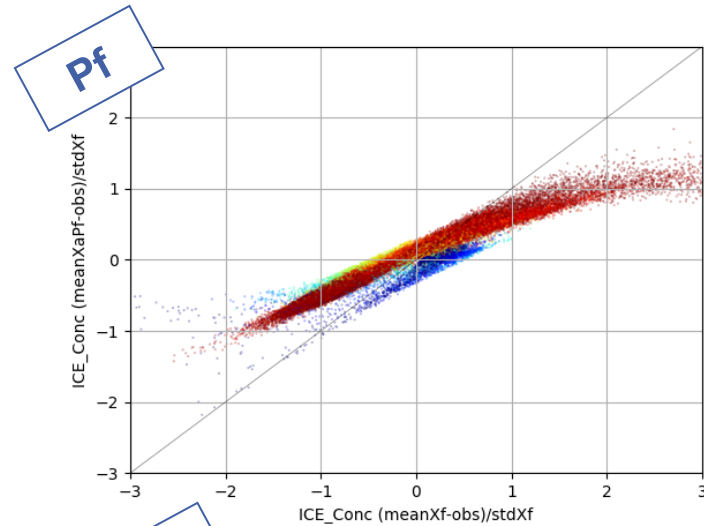
Realistic test case CREG4/NEMO3.6/LIM3

Reduced misfit of sample mean to nature run

Posterior vs Prior

Sea Ice Concentration for all the domain

Blue : low values of SIC
Red : High values of SIC



- **Sea-ice concentration priors are strongly non-Gaussian;**
- **Using very simple examples it was possible to demonstrate the benefit of using Gaussian Anamorphosis in the assimilation of sea-ice concentration**
- **In particular, Gaussian Anamorphosis seems to improve reliability and the representation of the posterior pdf;**
- **In a more realistic framework bias is an important component of the DA system, then observations tend to be outside the ensemble pack limiting the analysis correction;**
- **in addition the estimation of the observation error in the anamorphic space is not straightforward and could deteriorate the efficiency in case of wrong estimation.**
- **this study provides preliminary encouraging results but this work must be continued before an implementation in an operational system. In particular, tests of this method on a 3D analysis tools have not been done.**