



Data assimilation at NERSC

L. BERTINO, P. SAKOV, F. COUNILLON,
E. SIMON

NERSC, Bergen

SANGOMA KO meeting, Ulg, 25th Nov. 2011



With the support from the eVITA-EnKF project from the RCN
and the MyOcean project from the EU

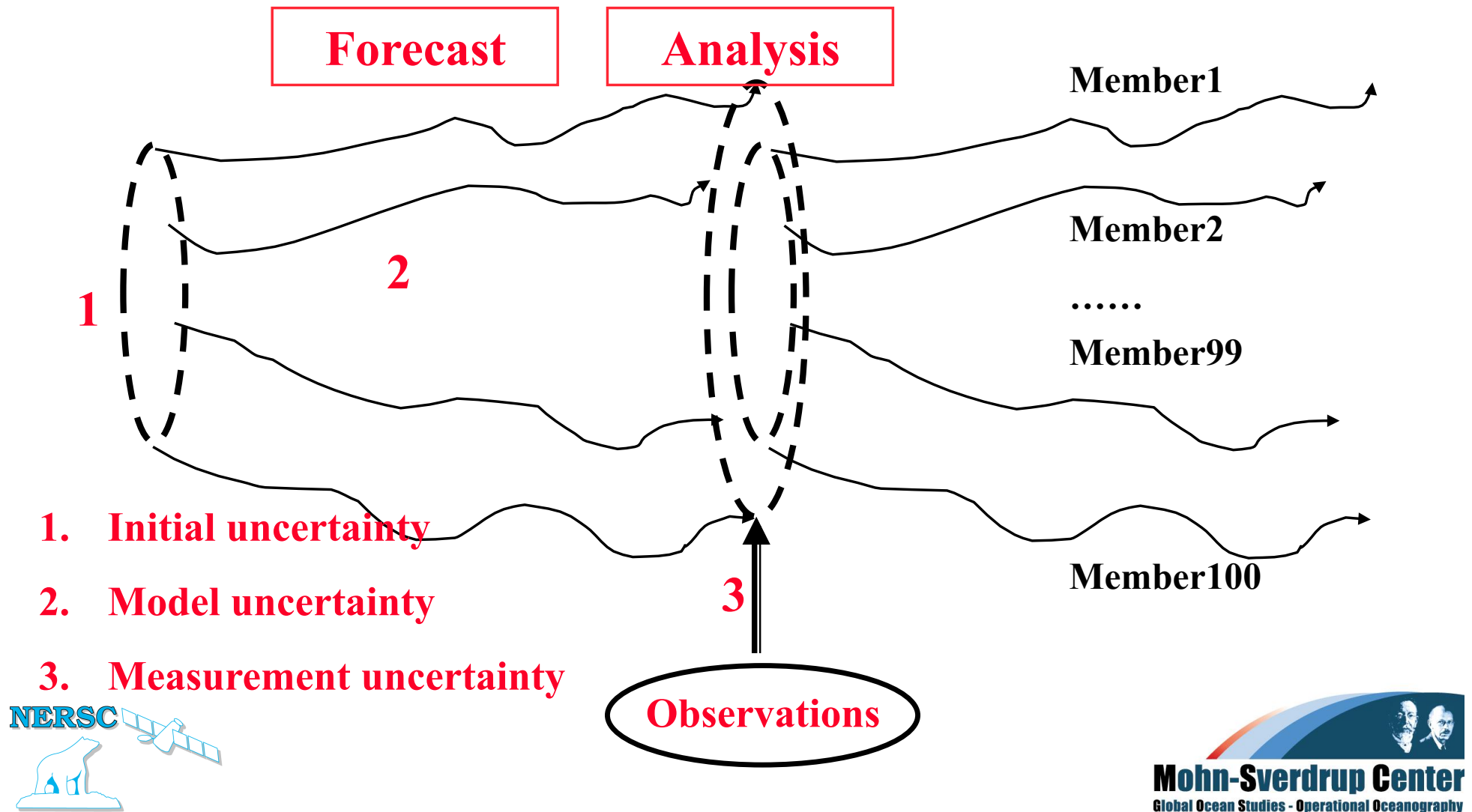


Outline

- Results from the TOPAZ pilot reanalysis (2003-2008)
- EnKF with Gaussian anamorphosis
- Iterated EnKF for highly non-linear models
- Relevant “Tier 1” R&D activities in MyOcean

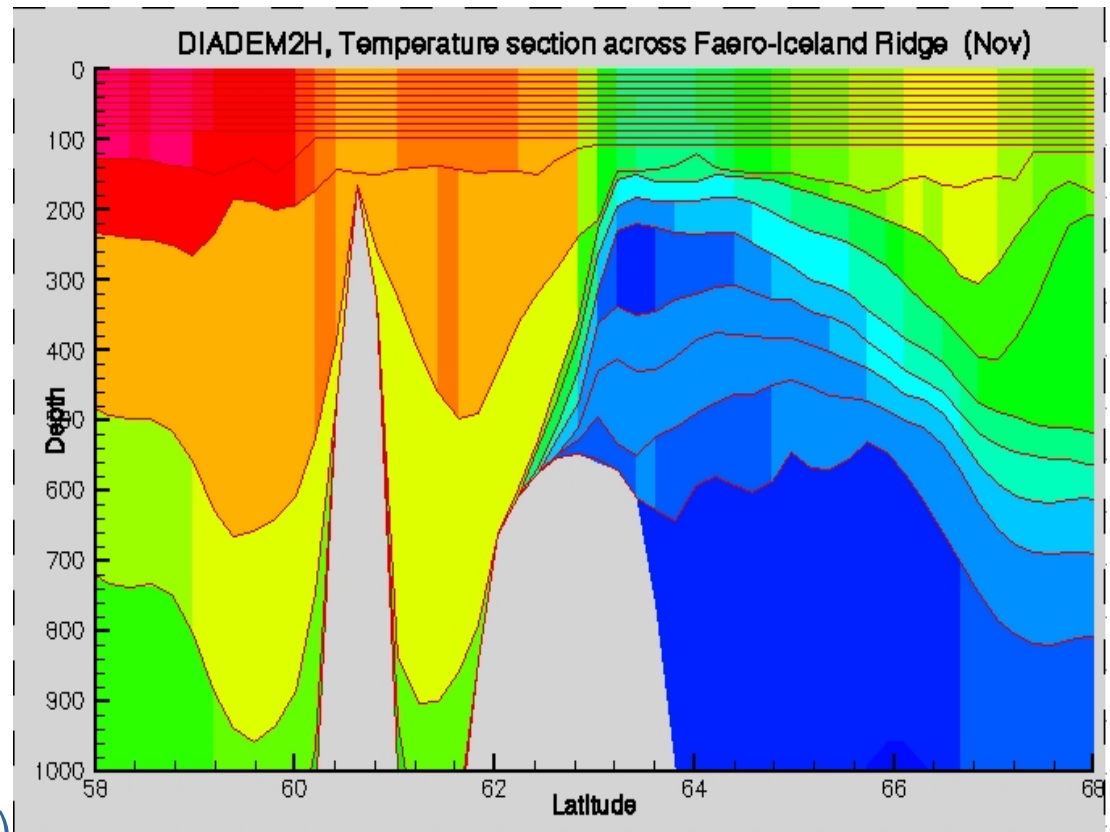


Ensemble Kalman filtering

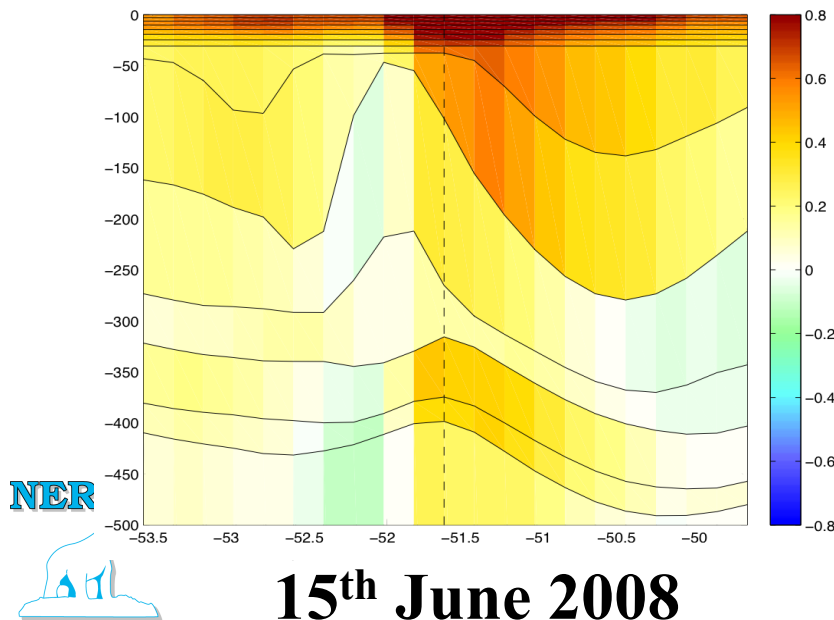
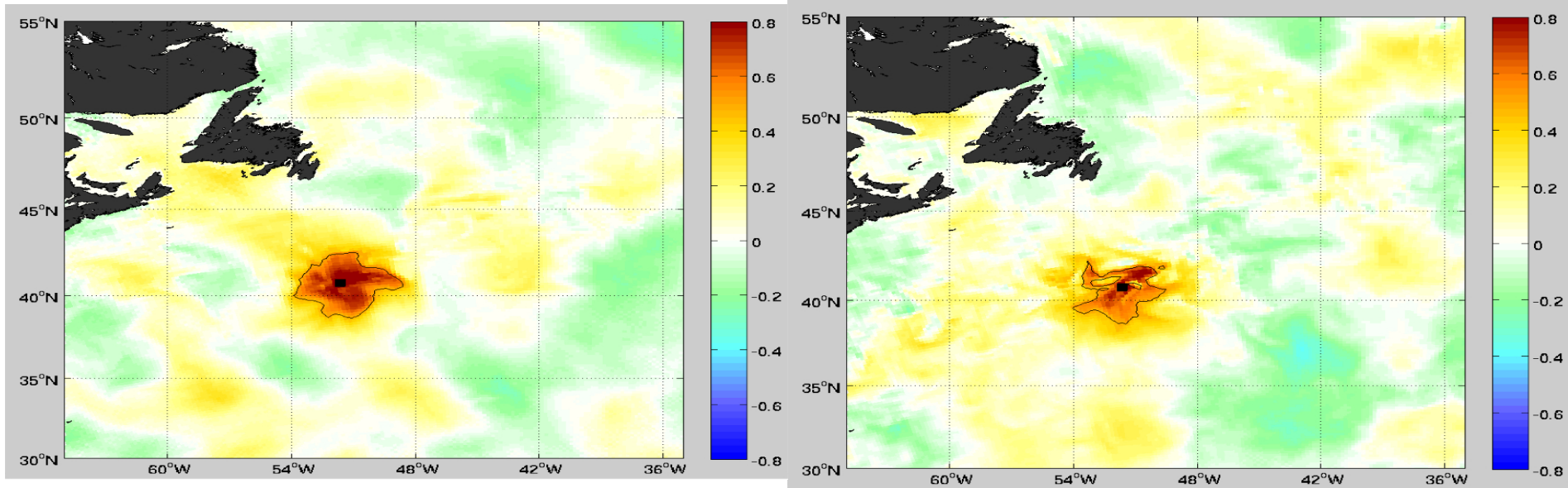


The HYCOM model at NERSC

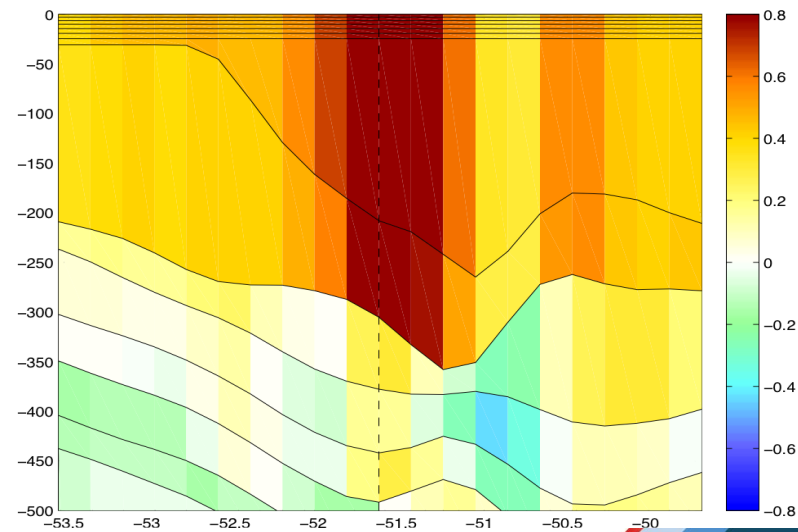
- 3D numerical ocean model
 - Hybrid Coordinate Ocean model, HYCOM (U. Miami)
- Hybrid vertical coordinate
 - Isopycnal in the interior
 - Z-coordinate at the surface
 - TOPAZ4 uses 28 layers
- Coupling to sea ice model
 - EVP dynamics
 - Semtner Thermodynamics
- Data assimilation: EnKF
 - 3D State variables (u,v,T,S,dp)
 - 2D State variables (ub,vb,pb,ice...)



EnKF Correlations, SST



15th June 2008



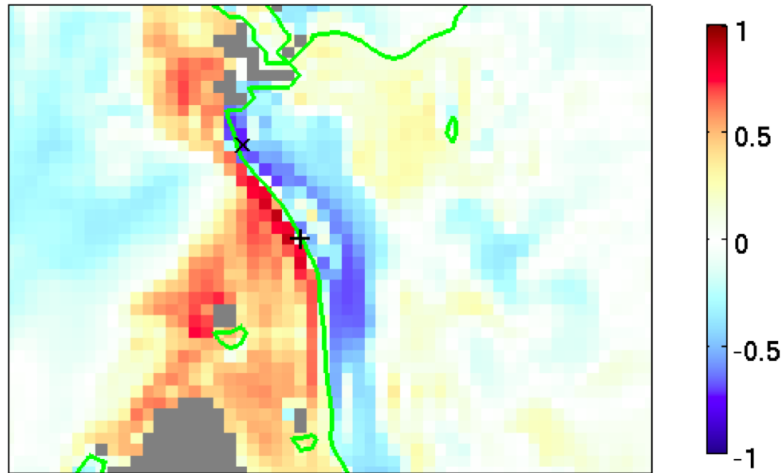
24 Dec. 2008 **Mohn-Sverdrup Center**
Global Ocean Studies - Operational Oceanography



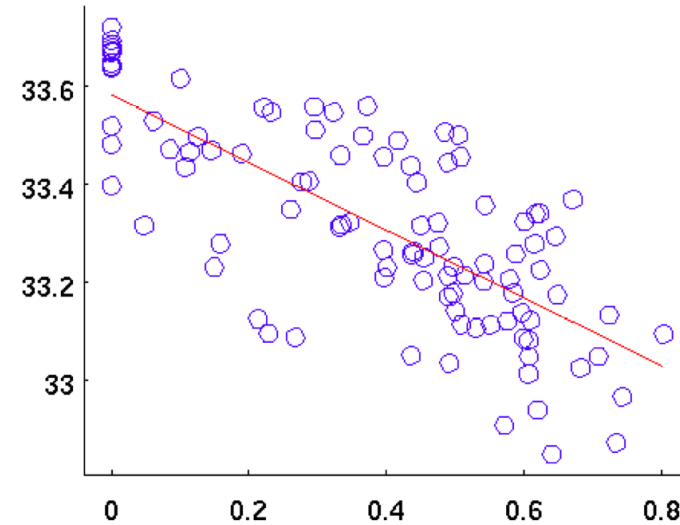
Why dynamic Data Assimilation in coupled ice-ocean model?

Dynamic ensemble

Correlation between ICEC and SSS



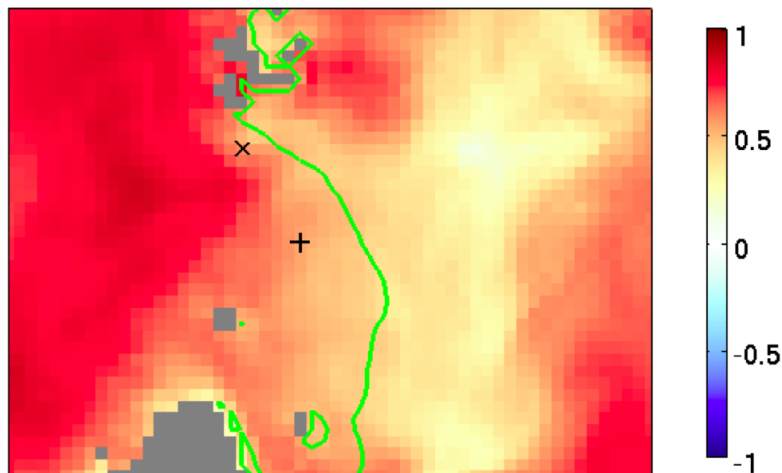
Scattergram between ICEC and SSS



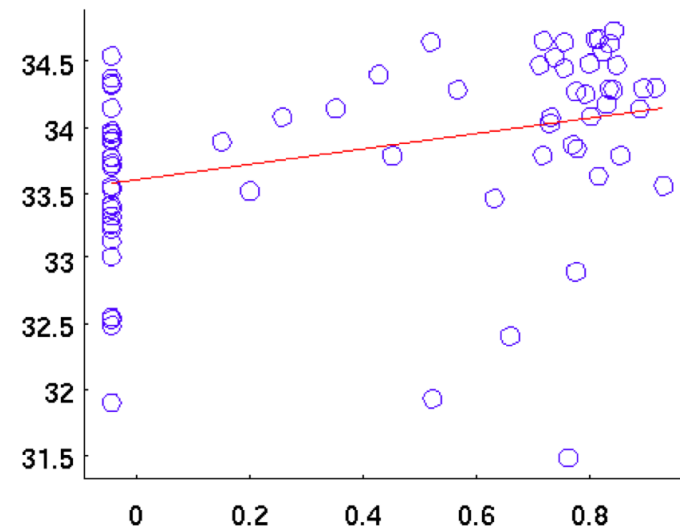
Also see *Lisæter et al. Oc. Dyn. 2003*

Static ensemble

Correlation between ICEC and SSS



Scattergram between ICEC and SSS



Conservation of properties

Evensen (2003)

- Update equation

$$X_n^a = X_n^f + K_n (Y_n - H(X_n^f))$$

- Factorize by X_n^f

$$X_n^a = X_n^f \cdot T$$

T: Transform matrix (size 100 x 100), also sometimes called X_5

Kalman gain:

$$K_n = X_n^f X_n'^f T H^T \cdot$$

$$(H X_n'^f X_n'^f T H^T + R)^{-1}$$

The transform T ensures **conservation of linear properties** (geostrophic balance), but not the others.

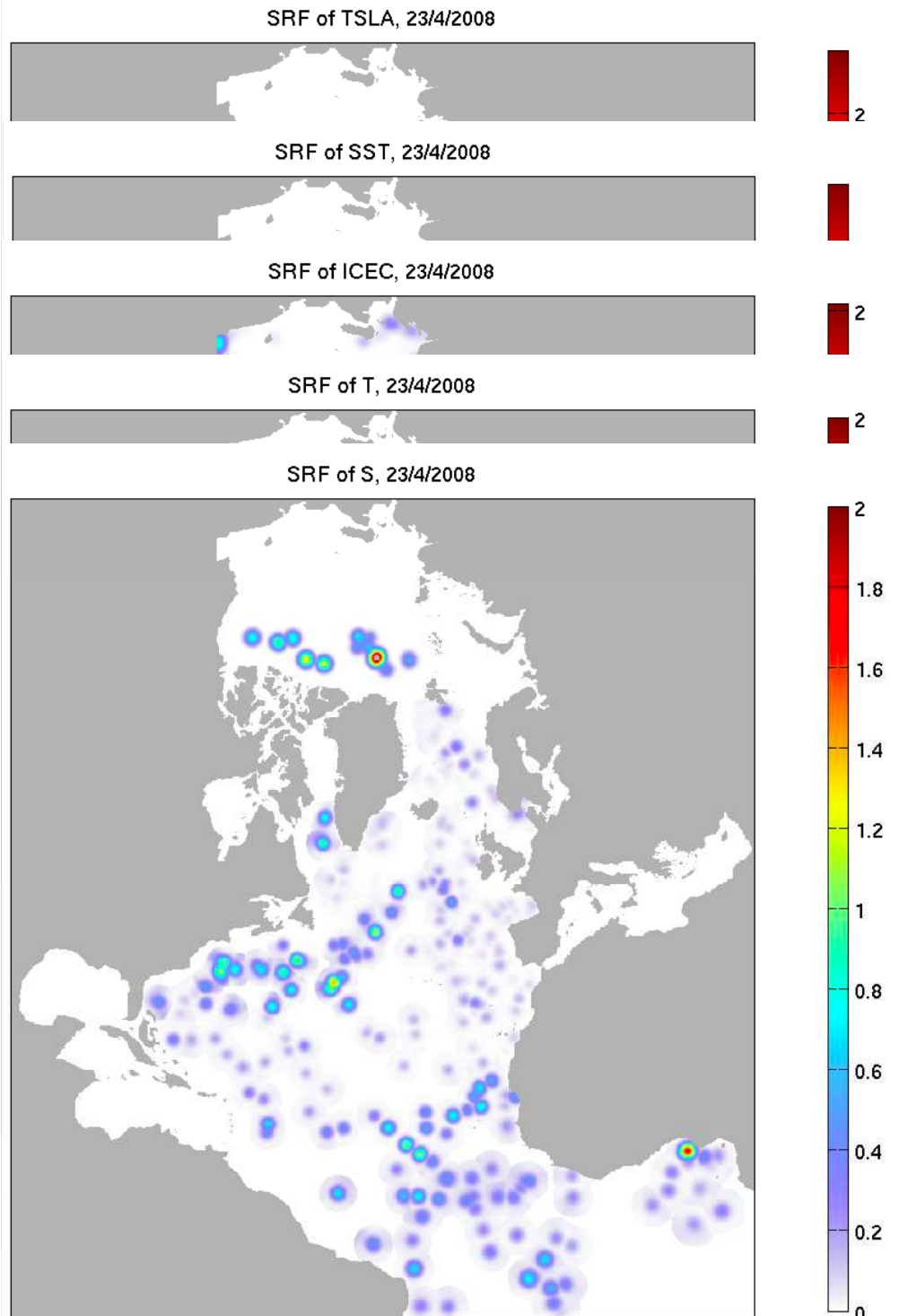


Ensemble X , anomalies $X' = X - \langle X \rangle$



The TOPAZ system

- DEnKF, **asynchronous**
 - 100 members
 - Local analysis (~90 km radius)
- Model state:
 - 3D variables (u,v, T, S, d)
 - 2D variables (ice, ...)
 - 800x880x148 = **104 million** variables
- Observations:
 - **Sea Level Anomalies (CLS)**
 - SST (NOAA, then OSTIA)
 - Sea Ice Concentr. (AMSR, NSIDC)
 - **Sea ice drift (CERSAT)**
 - T/S profiles (Argo, ITPs, field exps.)
 - **400,000 observations** per week



Computations

DEnKF 100 members

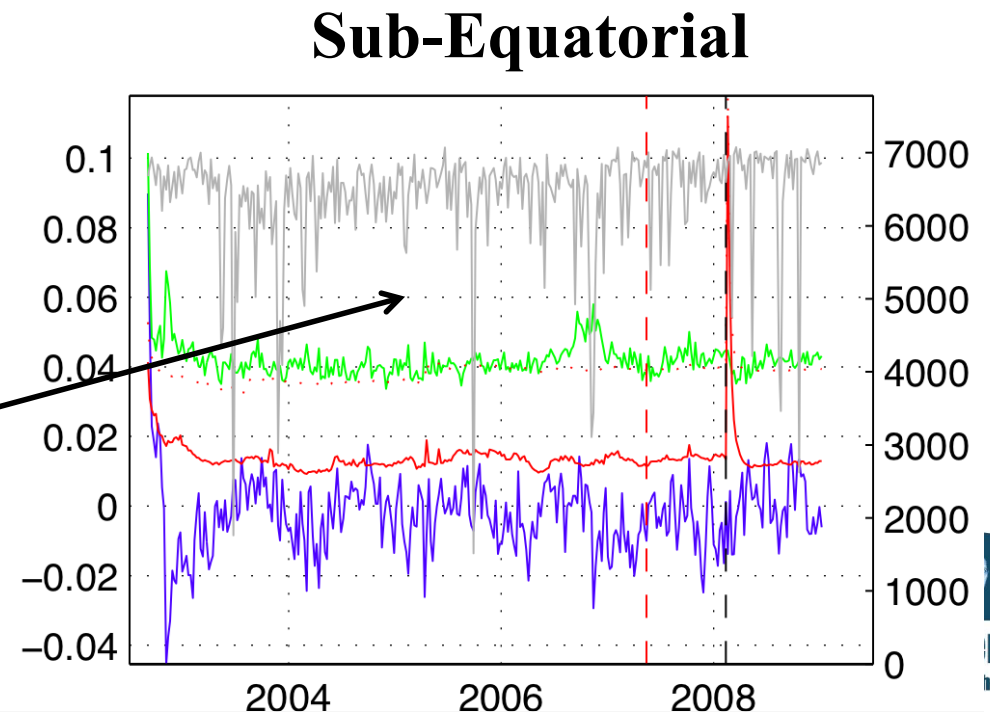
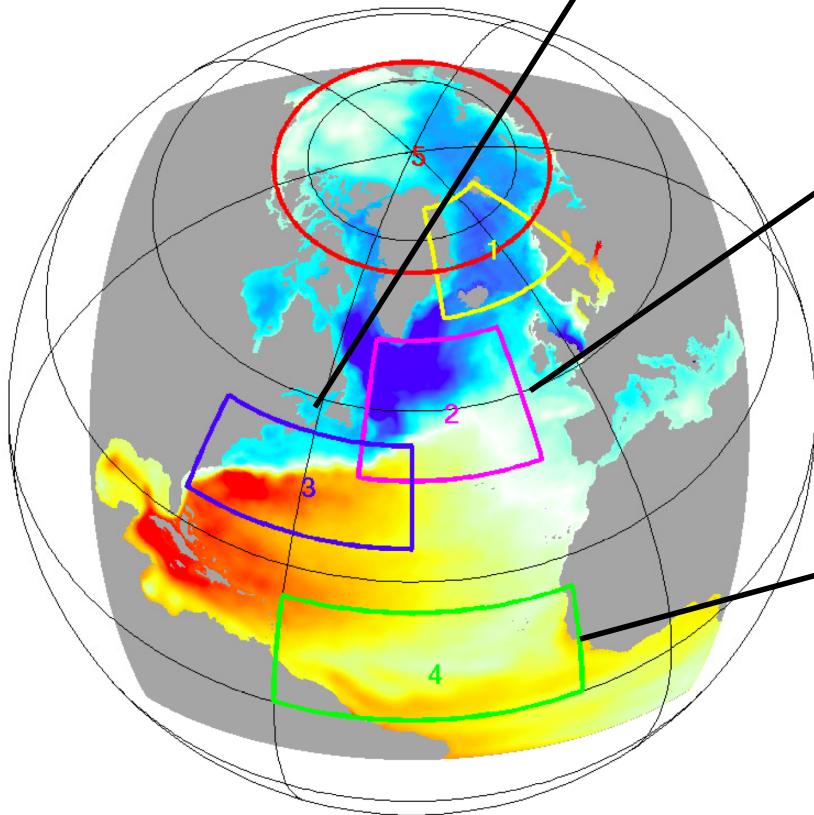
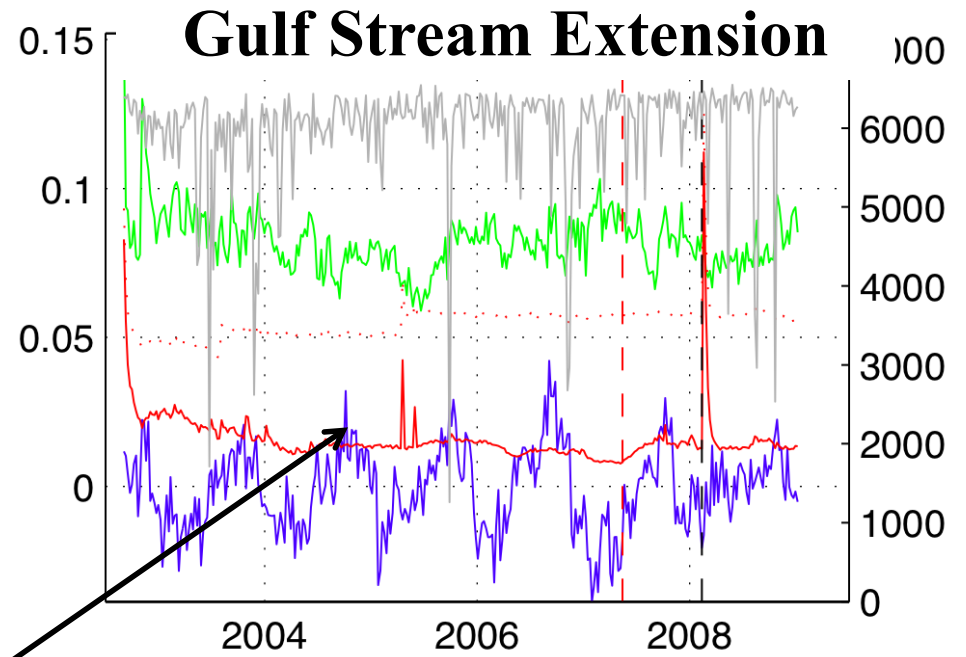
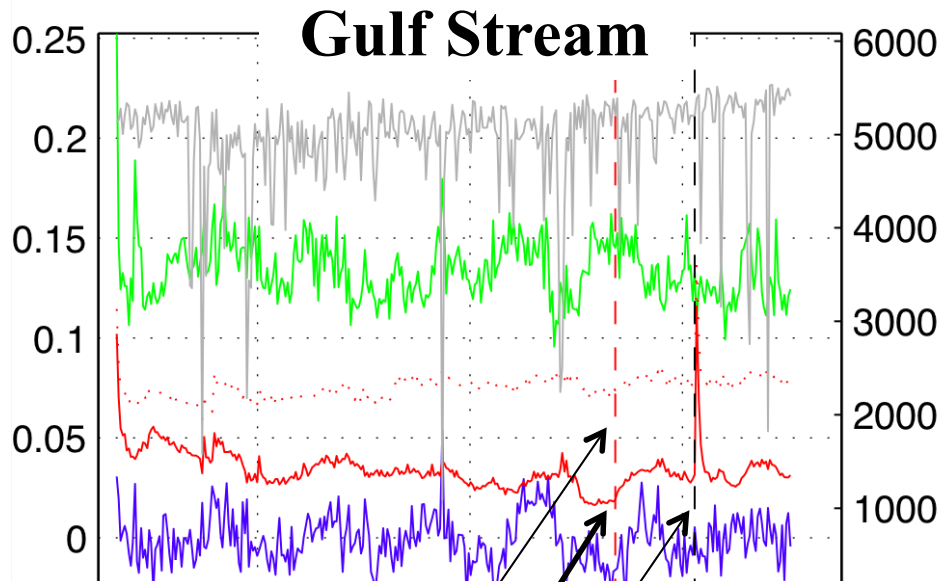
- Ensemble Forecast
- 2500 CPU hours / cycle
- Embarrassingly parallel
- 100x 133 CPU 11 min jobs
- Each job requires 400 Mb
 - MPI parallelization
- Analysis
- 20 CPU hours / update
- 6 datasets simultaneously
- One 20 CPU 1h job
- Memory required 1 Gb
 - MPI parallelization
- HPC Machine:
- Cray XT4, Installed 2008
 - 5500 CPUs, 55 Tflop/s
 - 1375 nodes (quad-core)
 - 1-4 Gb per node



Avoiding ensemble collapse?

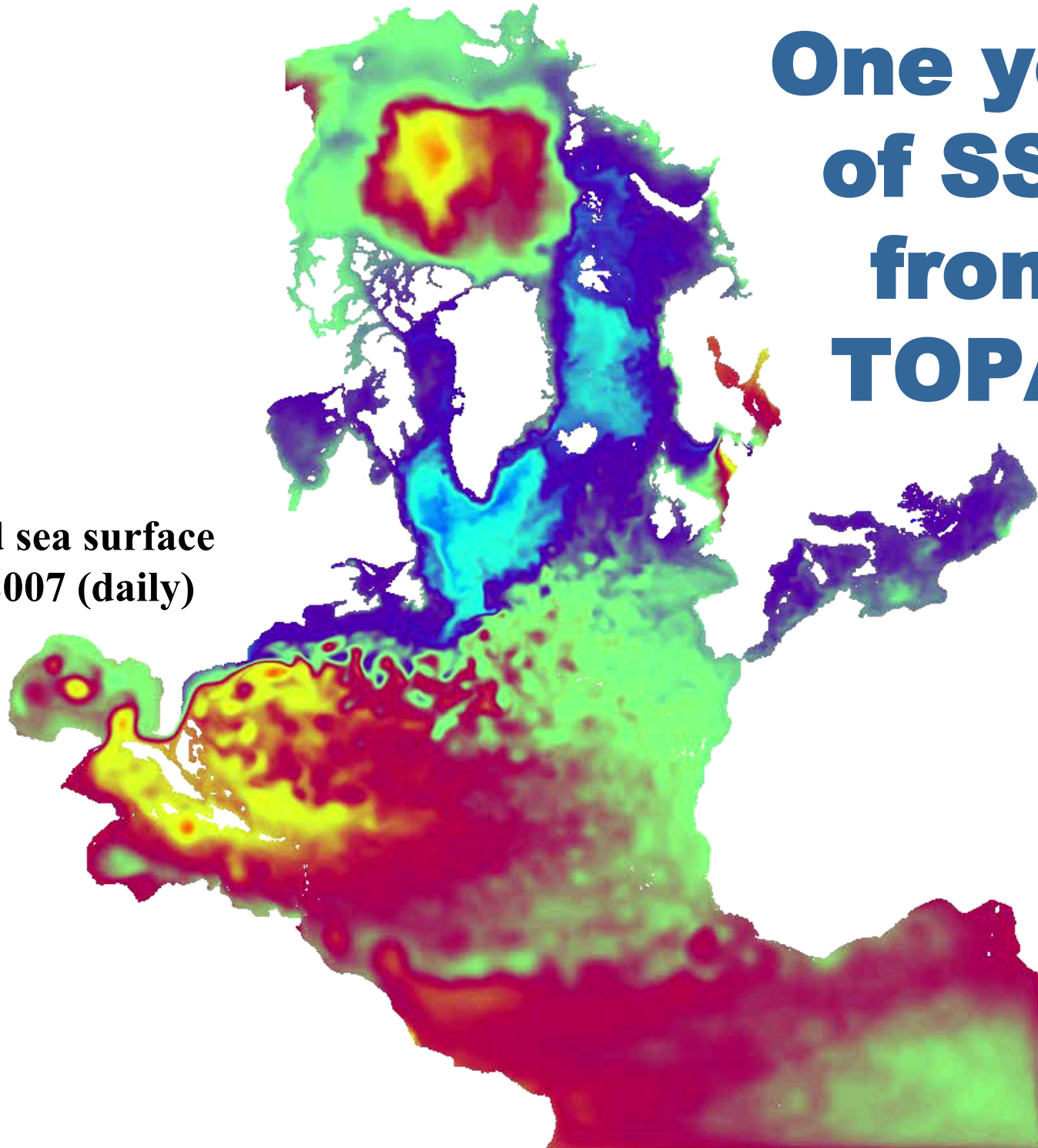
- Initial error
 - Interannual variability + ensemble run with model errors
- Model errors:
 - Winds, air temperature,
 - e-minus-p, cloud cover,
 - Static parameters:
 - mean SSH, mean SST, sea ice rheology parameters (stress tensor)
- Remediation of EnKF shortcomings:
 - Inflation: 1%
 - *Moderation of observations*
 - *Adaptive pre-screening of observations (if pdfs do not match: stretch!)*
 - *Exaggerate obs errors x2 for anomalies update only*





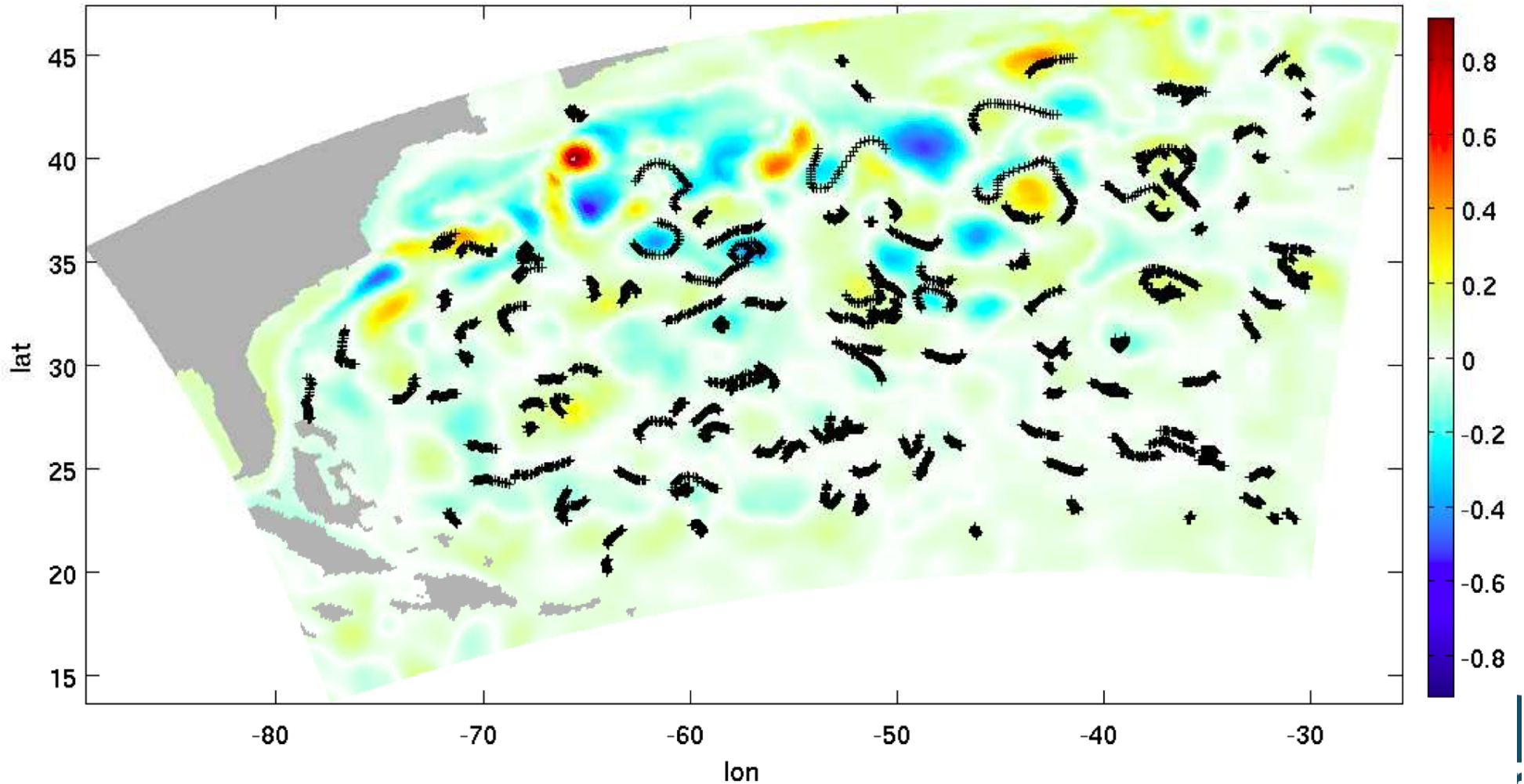
One year of SSH from TOPAZ

Reanalyzed sea surface
heights in 2007 (daily)



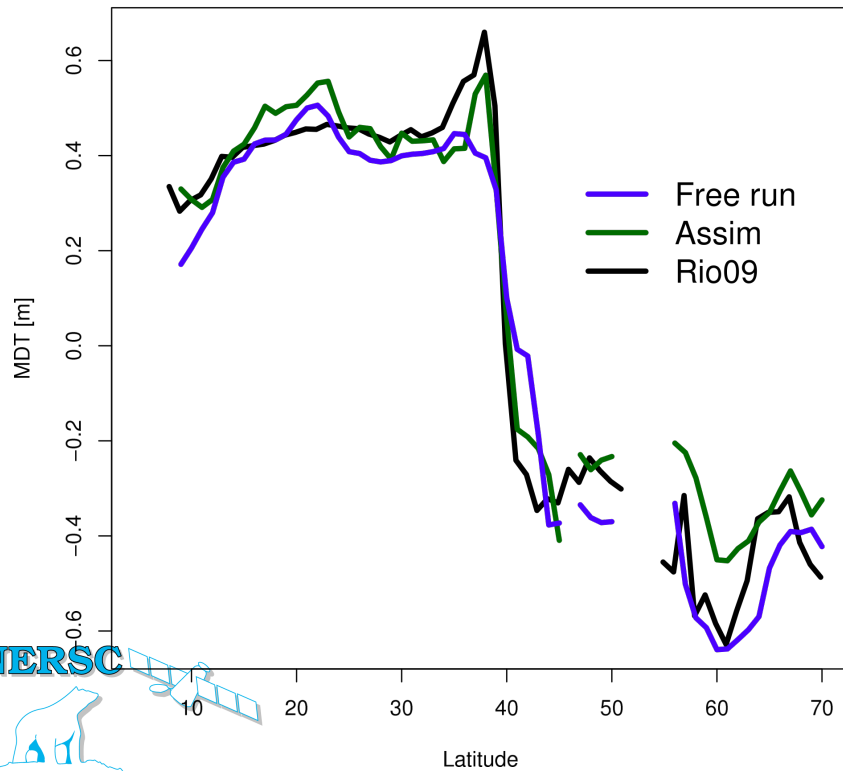
Independent data: surface drifters

9 January 2008: SLA from TOPAZ reanalysis + drifters (± 4 days)

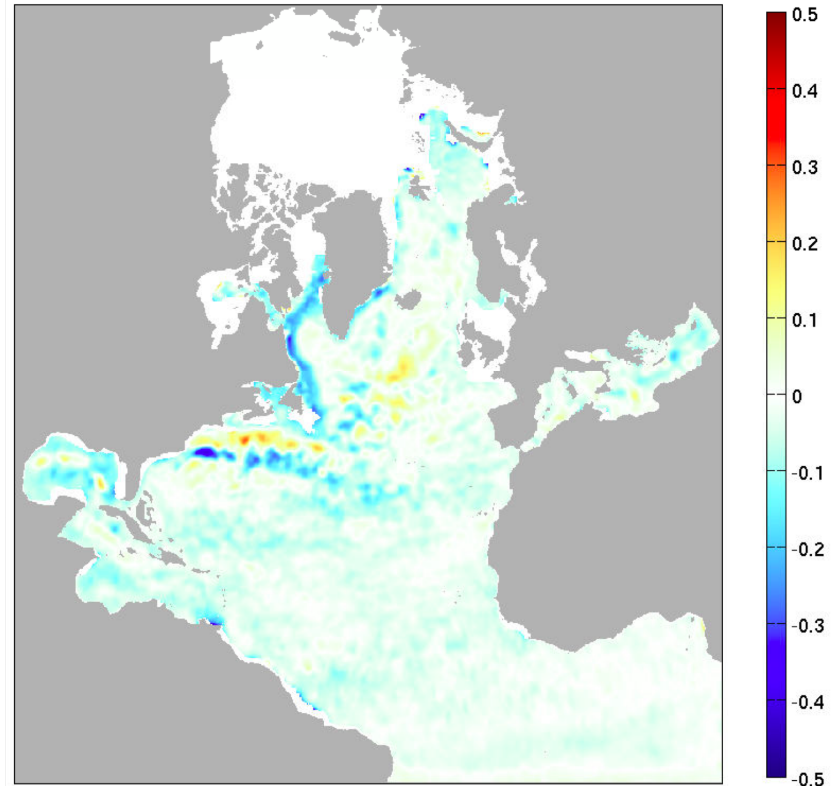


Bias estimation

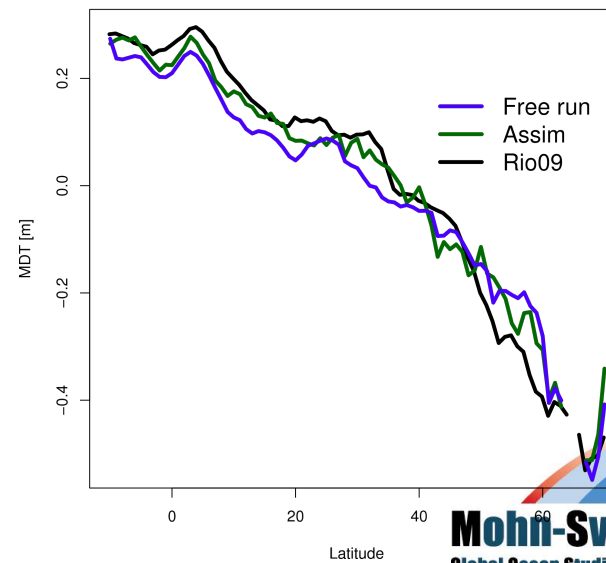
Section at 60W



MSSH bias on 7 January 2009

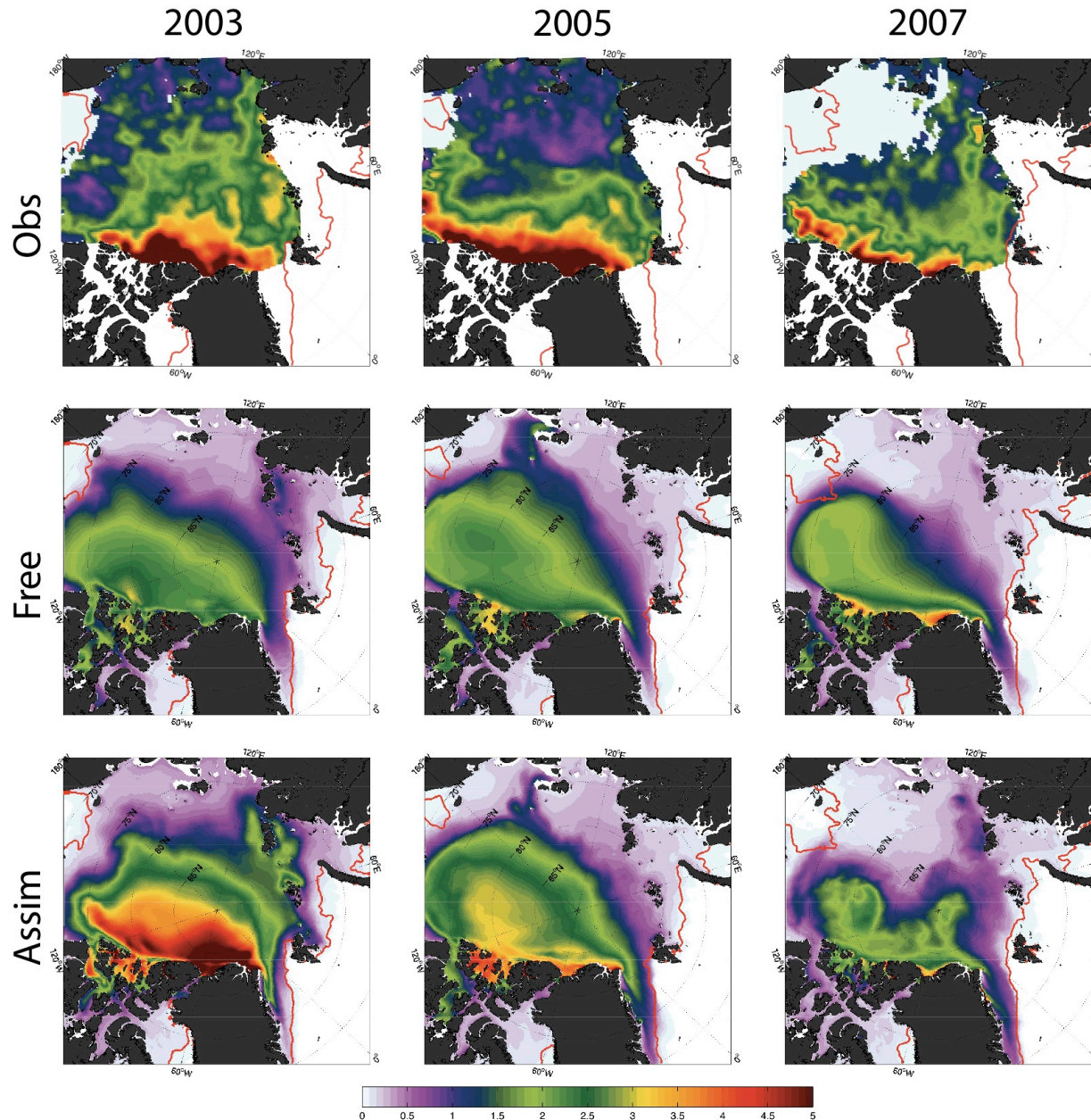


Section at 20W

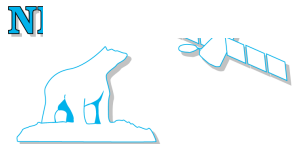
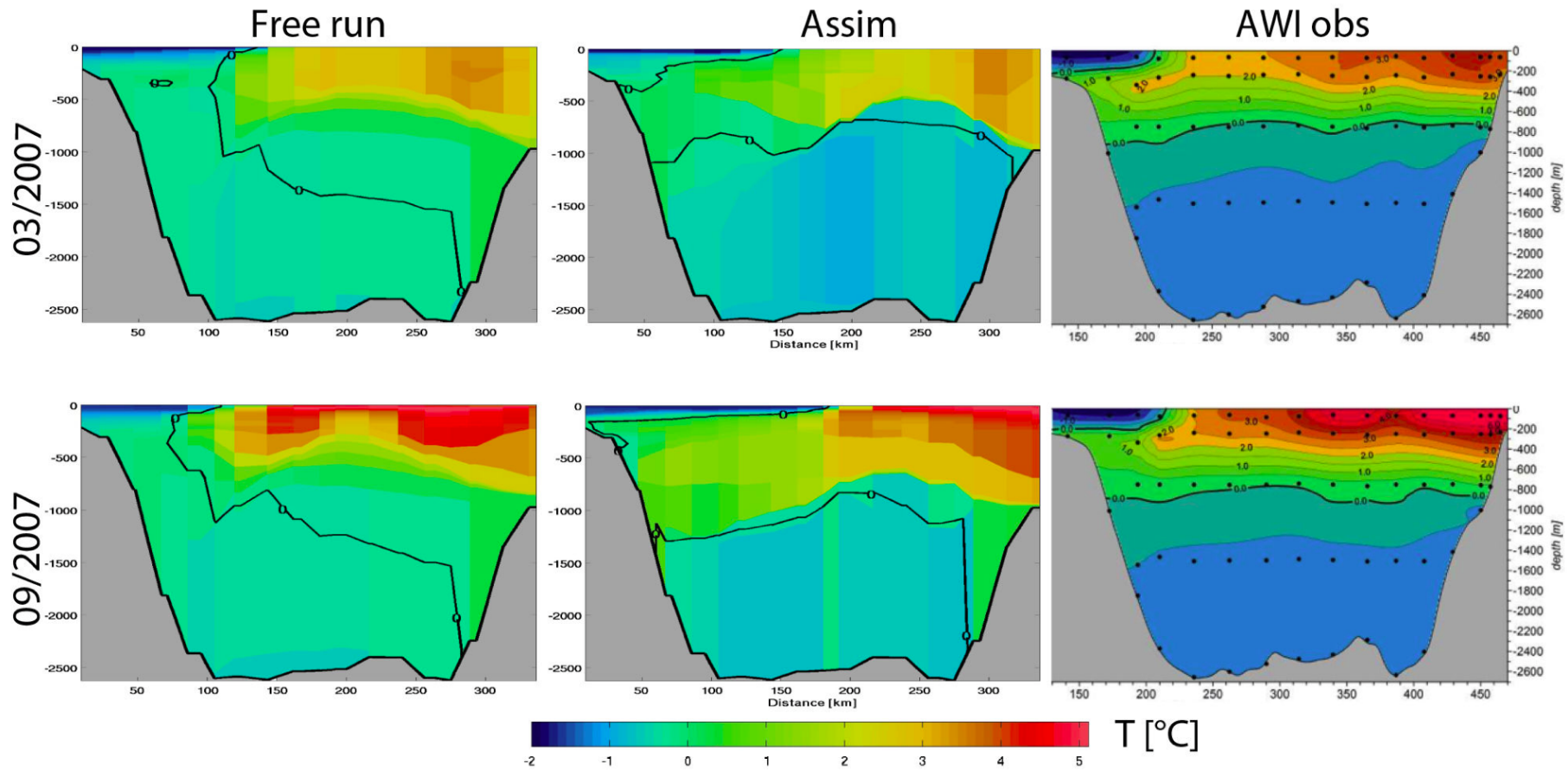


Not observed: Ice thickness

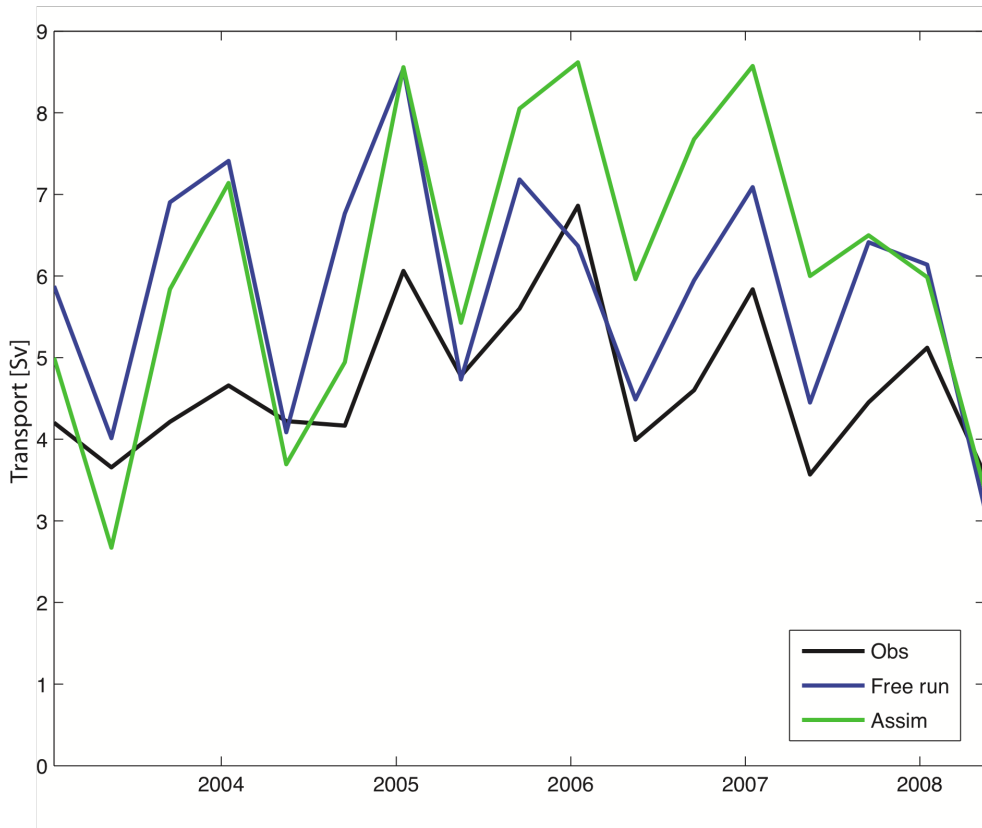
Independent
satellite data



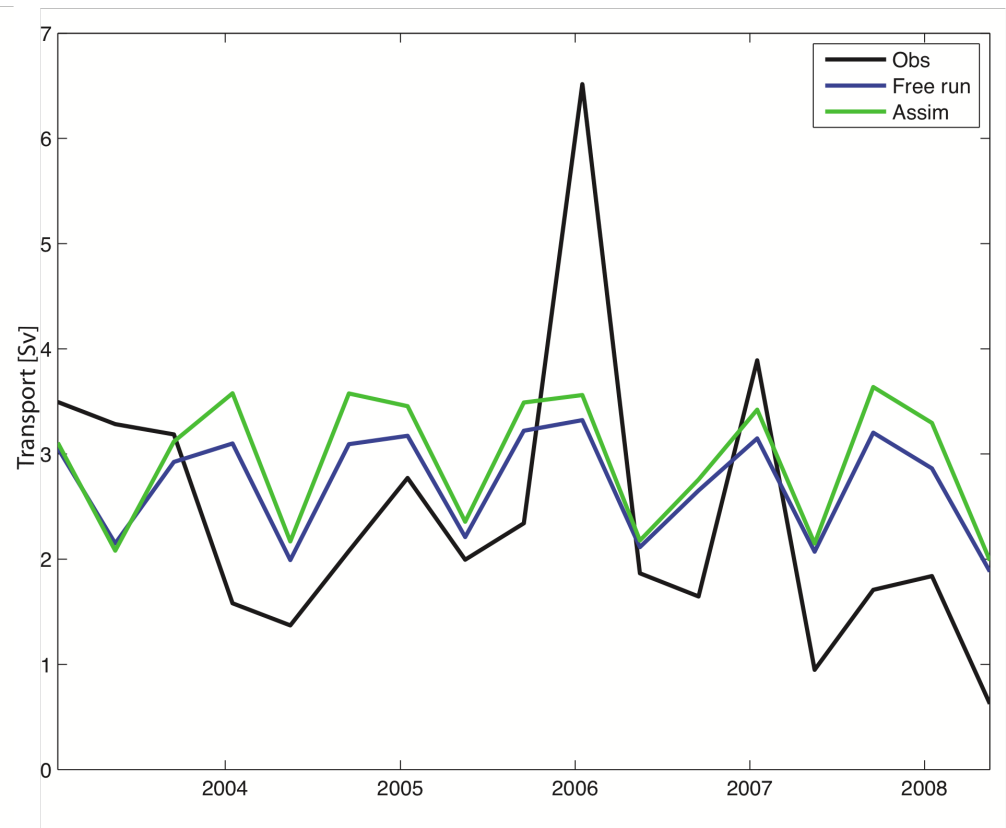
Oceanographic Validation



Water Transport across sections



**Svinøy Section
Net flux**



**Barents Sea Opening
Net flux**



Conclusion

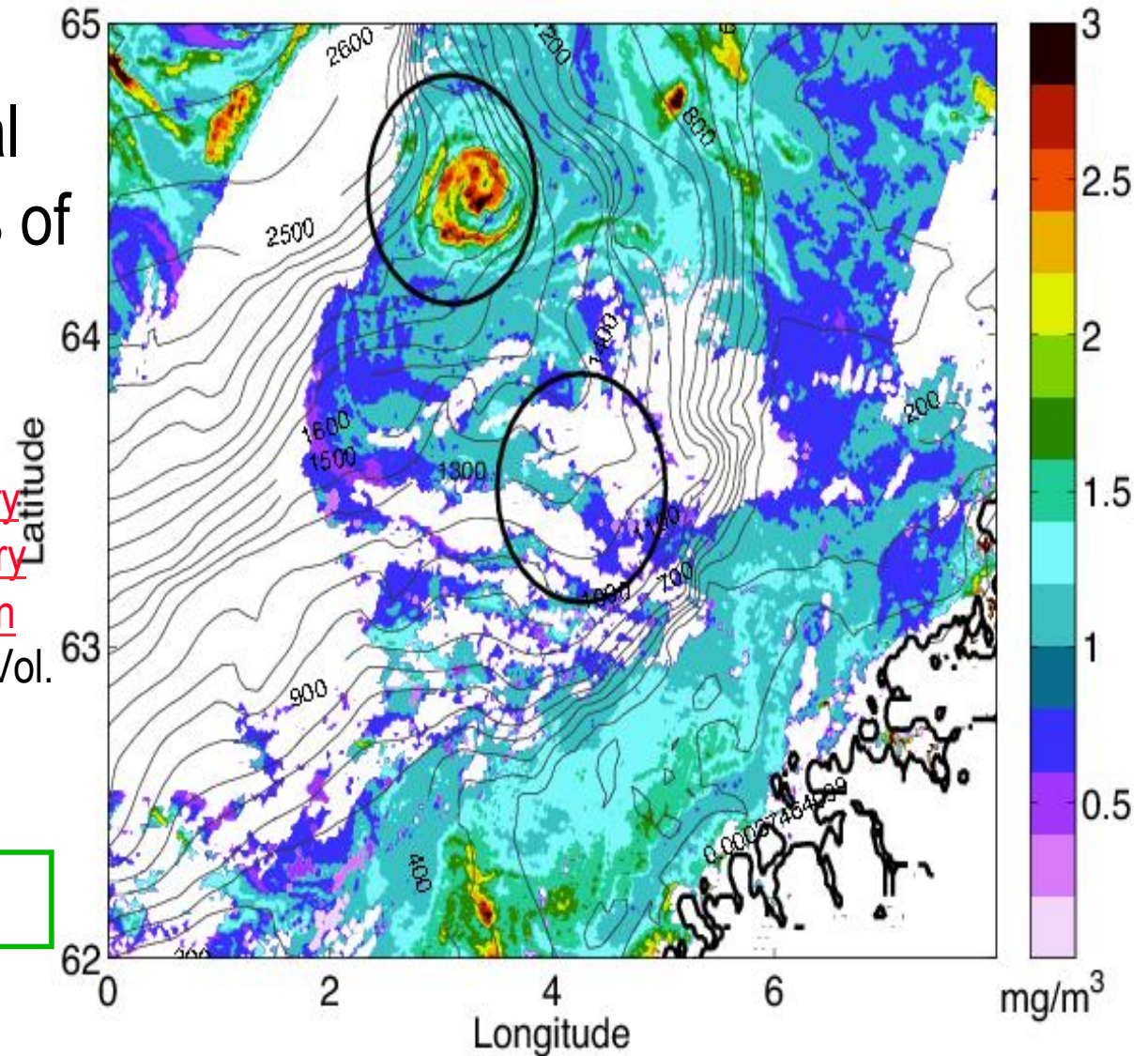
- Reanalysis publicly available on <http://topaz.nersc.no> and <http://myocean.eu.org>
- Code on <https://svn.nersc.no/enkf>
- The TOPAZ4 system is running a 20-years coupled ice-ocean reanalysis (ongoing)
 - No ensemble collapse or innovation drift
 - No assimilation “shocks” in transport time series
- Allows assimilation of various data types
 - Based on Monte-Carlo framework
 - Identified sources of error (Bayesian philosophy)
 - Fully multivariate method



Weather IN the oceans: Mesoscale physical-ecosystem interactions

MERIS ocean colour data

- Eddies can make local deserts or local oases of marine life
- Hansen, C., A. Samuelsen:
[Influence of horizontal model grid resolution on the simulated primary production in an embedded primary production model in the Norwegian Sea](#): Journal of Marine Systems. Vol. 75, Issue: 1-2, pp 236-244, 2009.

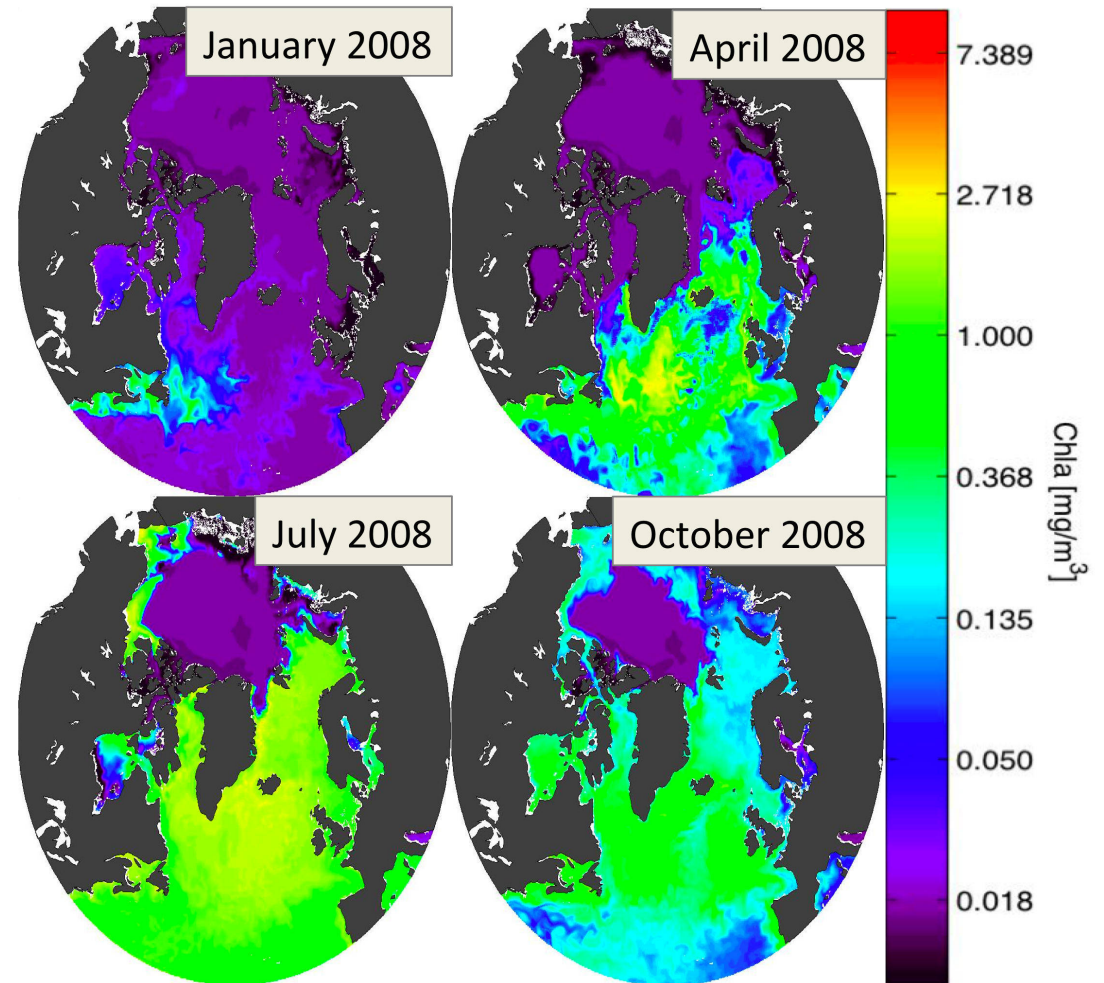


C. Hansen



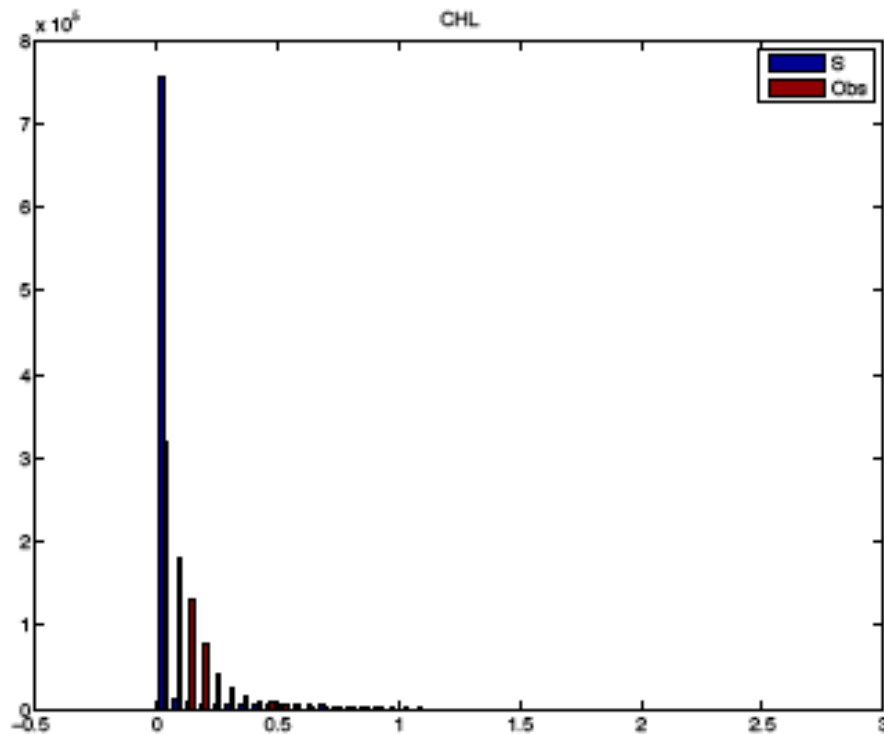
Ocean ecosystem anamorphosis – *EnKF* (E. Simon)

- Data:
 - Satellite, ocean colour
 - SeaWiFS
- Problem
 - Coupled 3-dimensional physical-biological model
 - High-dimensional
 - Non-Gaussian variables

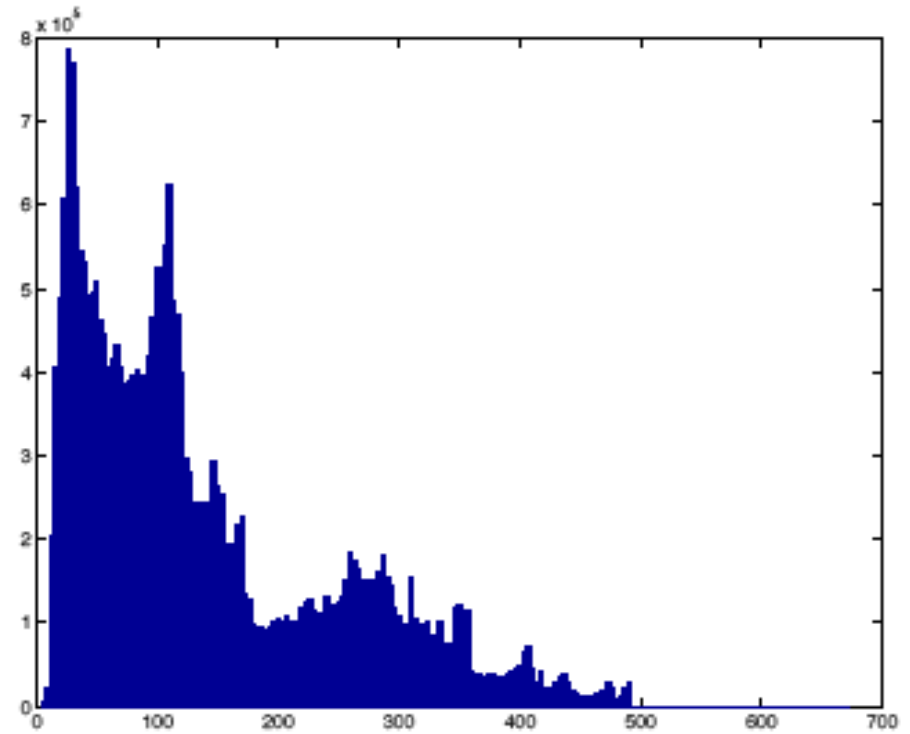


Non-Gaussian variables

Chlorophyll



Nitrate nutrients



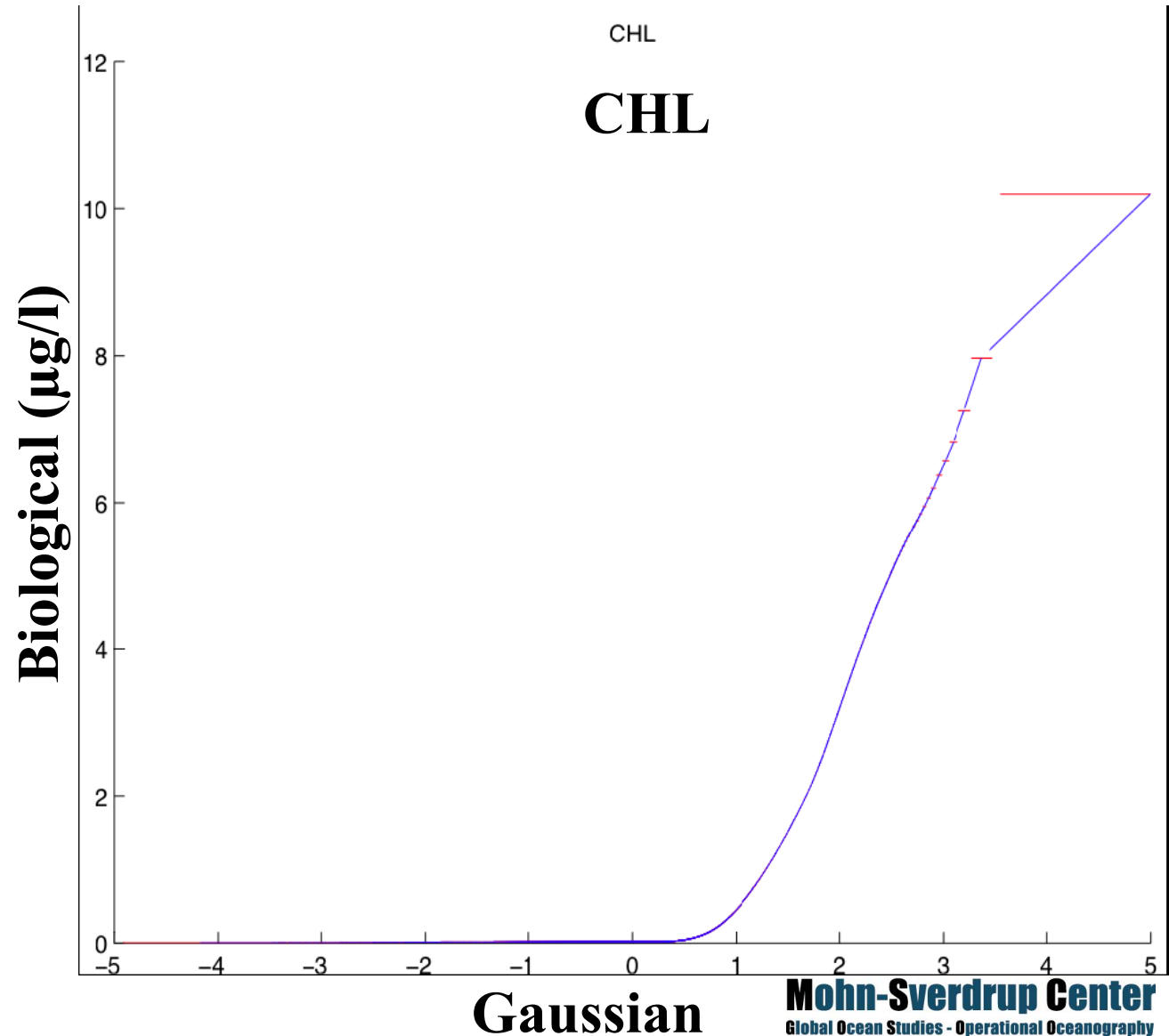
- Asymmetric distributions (right skewed)



- Positive valued

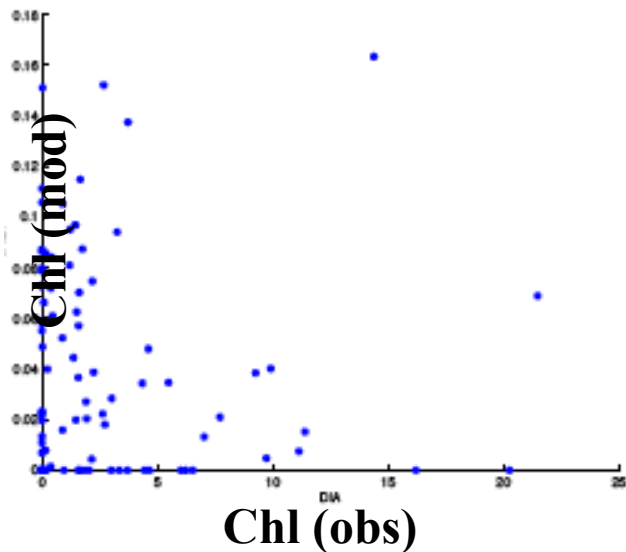
Gaussian anamorphosis

- Univariate transform
 - to Gaussian.
- More adequate for linear analysis
- Idea taken from Geostatistics

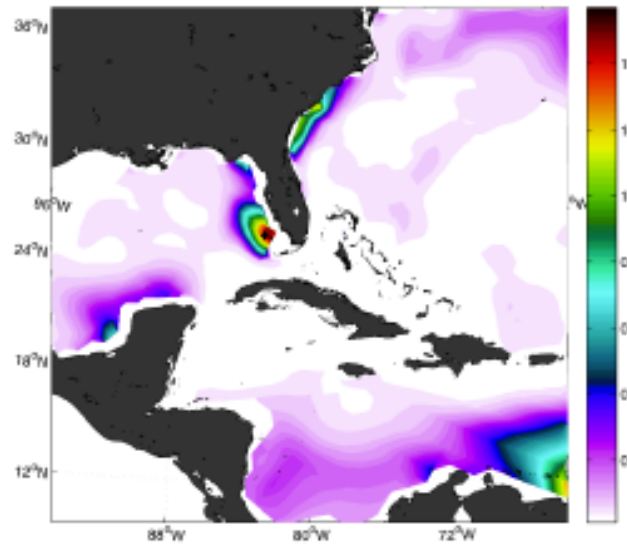


Why is Gaussianity important?

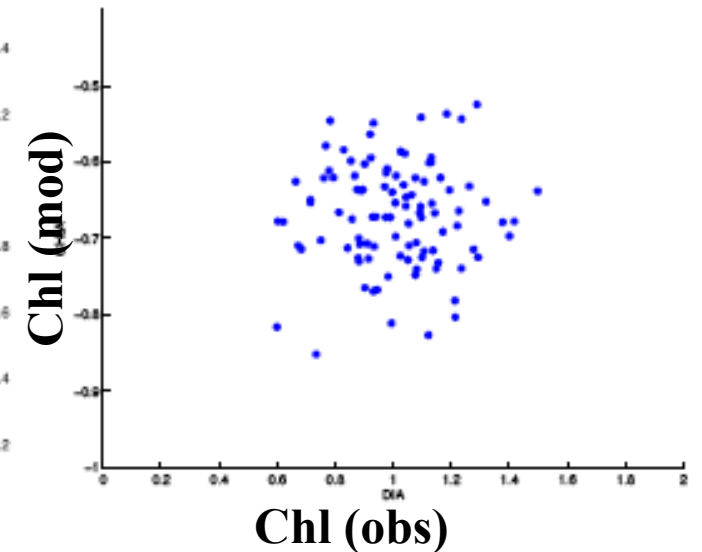
Standard



ECO : Mean of the forecast



ANA



- Makes scatterplots more linear
- Reduces the impact of outliers



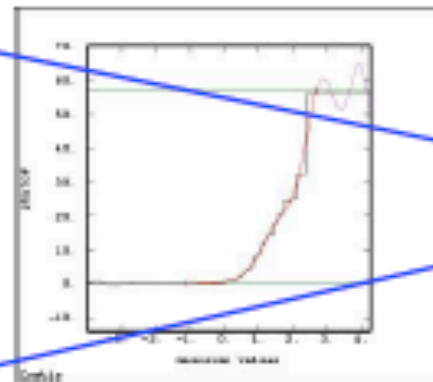
Gaussian Anamorphosis

Physical
operations:

Forecast
 $A_n^f = f(A_{n-1}^a)$

Forecast
 $A_{n+1}^f = f(A_n^a)$

Anamorphosis
function



Statistical
operation: **A** and **Y**
transformed

Analysis
 $A_n^a = A_n^f + K_n(Y_n - HA_n^f)$

[Bertino et al. 2003]

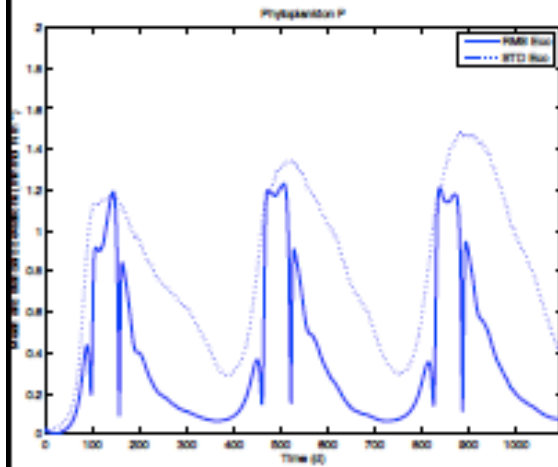
[Simon et al. 2009]

Estimating *loss to carnivores*

Biased g : parameters

Simon & Bertino, JMS 2011

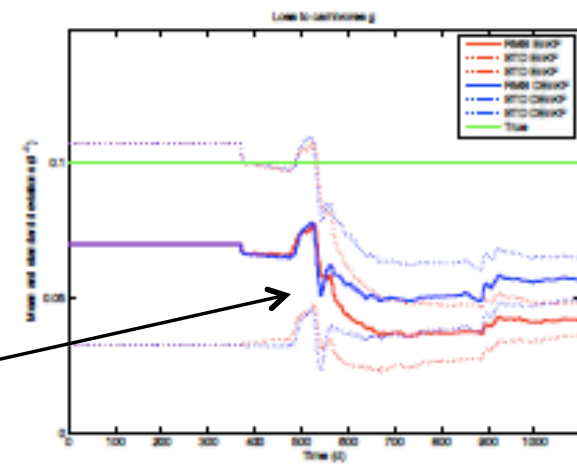
Errors on phytoplankton



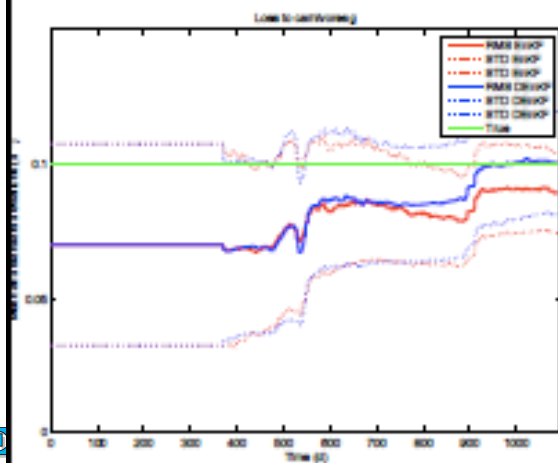
Biased loss to carnivores (g):
mean and spread.

Wrong direction!

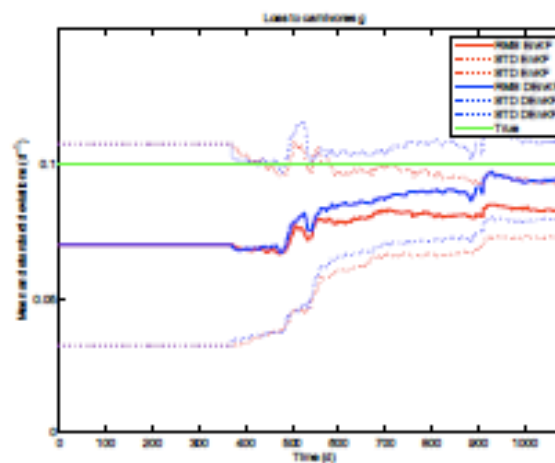
Without anamorphosis



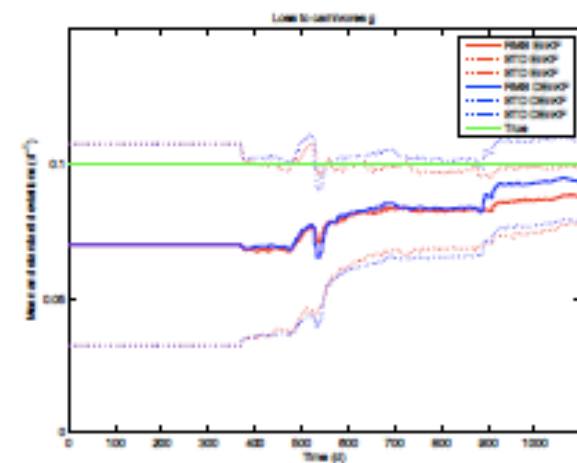
Static anamorphosis



Dynamical anamorphosis



Hybrid anamorphosis



- End of the 1st bloom: correction in the opposite direction without anam.
- Anamorphosis: slow convergence towards the true value (2 blooms).

The iterated EnKF for strongly non-linear systems

P. Sakov, D. Oliver, L. Bertino

Under final (?) review in MWR

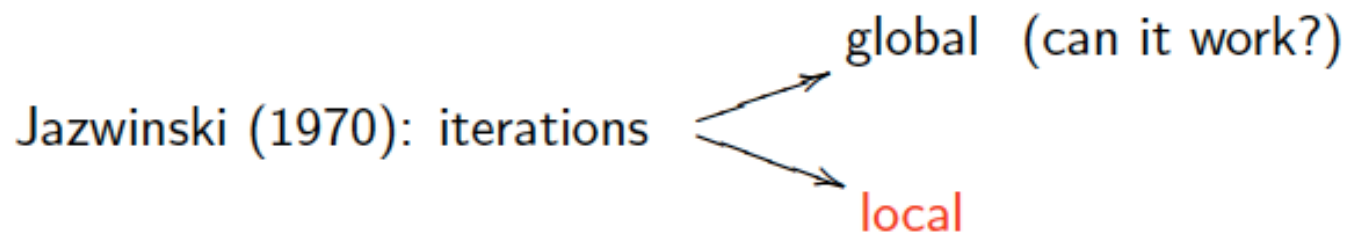


The concept

We consider:

- ▶ A nonlinear system with infrequent but rather precise observations
- ▶ So that propagation of the ensemble anomalies about the ensemble mean is substantially nonlinear (“strongly nonlinear system”)
- ▶ The system would be less nonlinear **after** assimilating future observations
- ▶ But the assimilation can not be done properly because the sensitivities are imprecise due to the nonlinearity (too big ensemble spread)

We seek solution through **iterations** of the linear solution



Iterative solution

The objective function: (perfect model framework)

$$\mathbf{x}_1^a = \arg \min_{\{\mathbf{x}_1\}} \left\{ (\mathbf{x}_1 - \mathbf{x}_1^f)^T (\mathbf{P}_1^f)^{-1} (\mathbf{x}_1 - \mathbf{x}_1^f) + [\mathbf{y}_2 - \mathcal{H}_2(\mathbf{x}_2)]^T (\mathbf{R}_2)^{-1} [\mathbf{y}_2 - \mathcal{H}_2(\mathbf{x}_2)] \right\}, \quad \mathbf{x}_2 = \mathcal{M}_{12}(\mathbf{x}_1)$$

Solution:

$$\mathbf{x}_1^{i+1} = \mathbf{x}_1^i + \mathbf{K}_{12}^i \left\{ \mathbf{y}_2 - \mathcal{H}_2 [\mathcal{M}_{12}(\mathbf{x}_1^i)] \right\} + \mathbf{P}_1^i (\mathbf{P}_1^f)^{-1} (\mathbf{x}_1^f - \mathbf{x}_1^i) \quad (\text{Tarantola, 2005, eq. 3.51}) \quad (1)$$

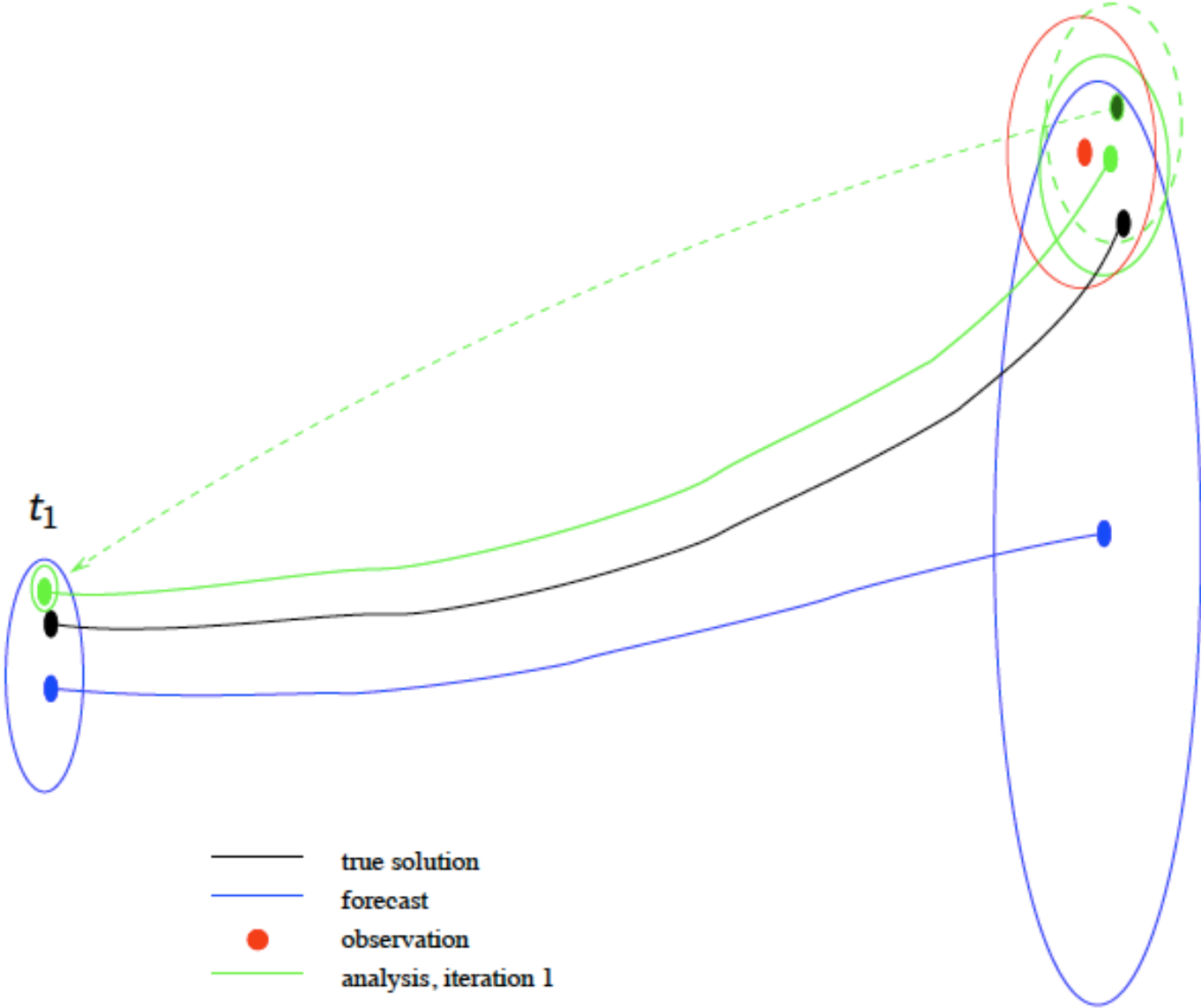
or

$$\mathbf{x}_1^{i+1} = \mathbf{x}_1^f + \mathbf{K}_{12}^i \left\{ \mathbf{y}_2 - \mathcal{H}_2 [\mathcal{M}_{12}(\mathbf{x}_1^i)] + \mathbf{H}_2^i \mathbf{M}_{12}^i (\mathbf{x}_1^i - \mathbf{x}_1^f) \right\} \quad (\text{Gu and Oliver, 2007, eqs. 12,13}) \quad (2)$$

where

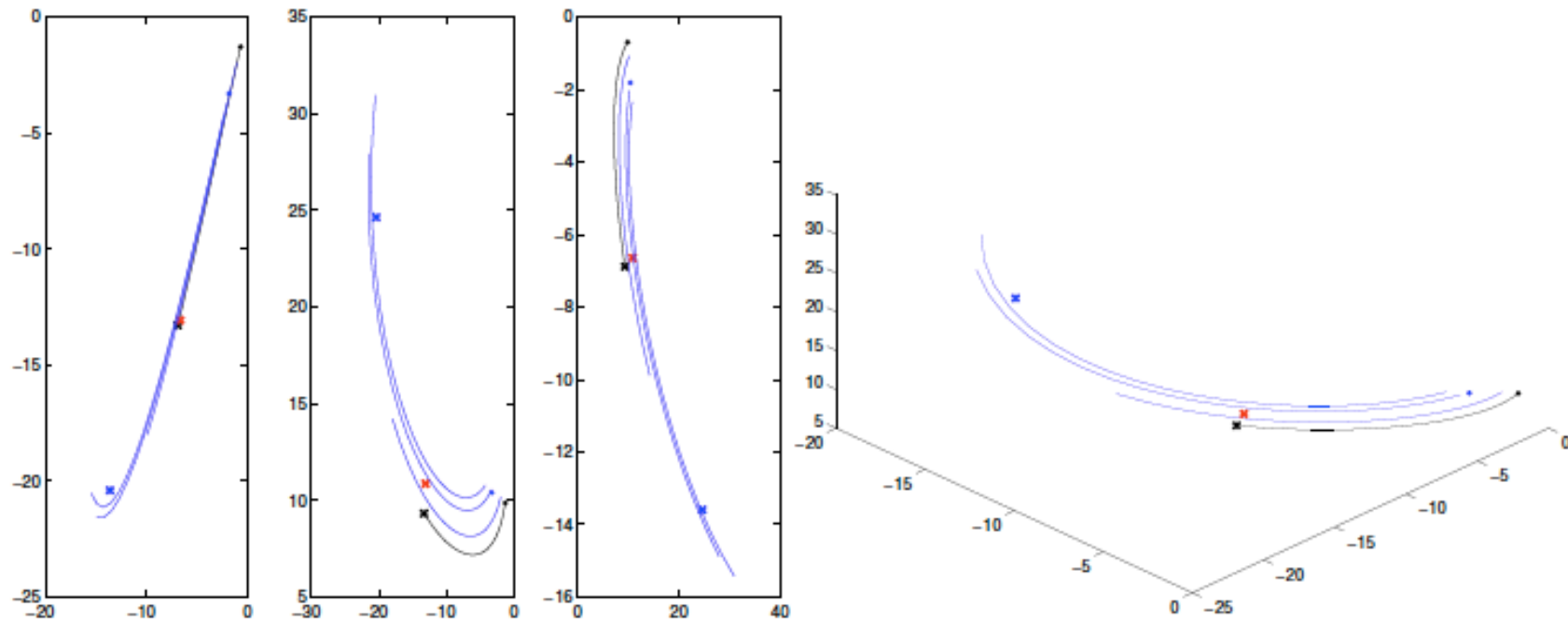
$$\mathbf{K}_{12}^i = \mathbf{P}_1^f (\mathbf{H}_2^i \mathbf{M}_{12}^i)^T \left[\mathbf{H}_2^i \mathbf{M}_{12}^i \mathbf{P}_1^f (\mathbf{H}_2^i \mathbf{M}_{12}^i)^T + \mathbf{R}_2 \right]^{-1},$$
$$\mathbf{P}_1^i = [\mathbf{I} - \mathbf{K}_{12}^i \mathbf{H}_2^i \mathbf{M}_{12}^i] \mathbf{P}_1^f$$

The Iterative EnKF



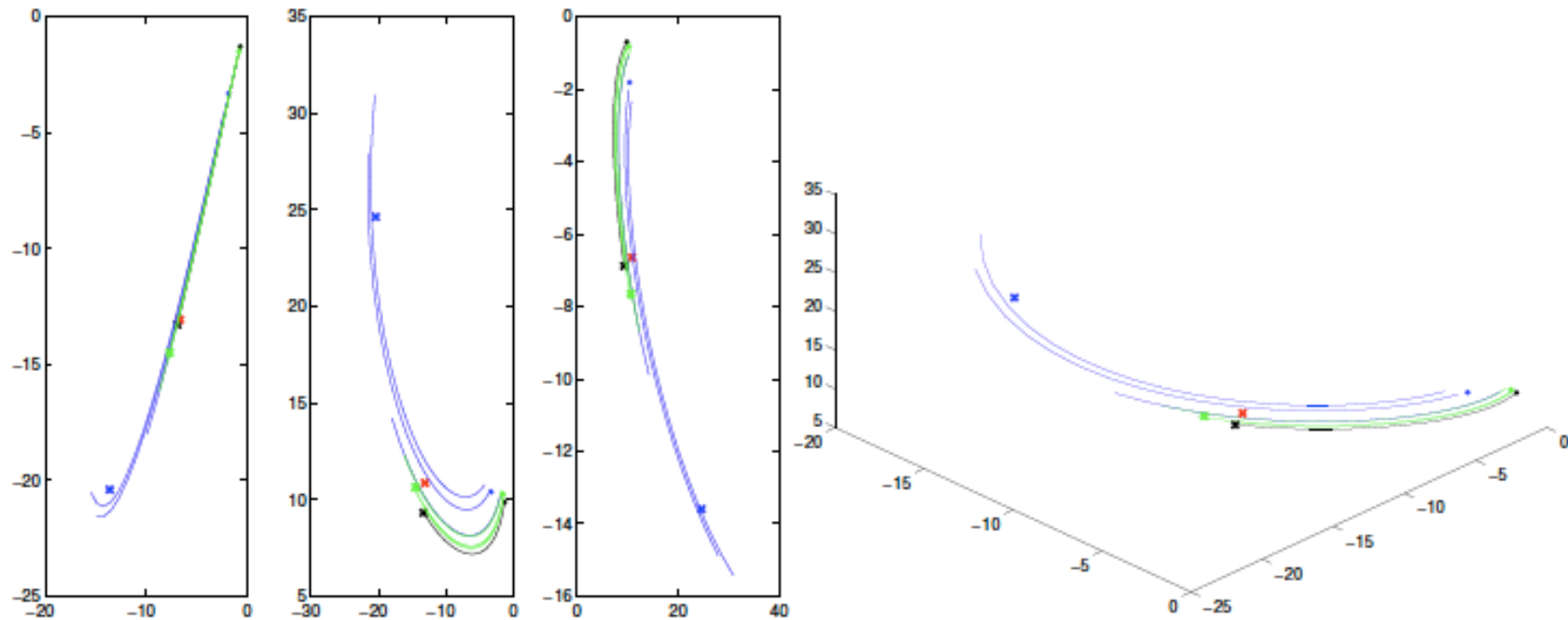
Example

Cycle 10 (1):



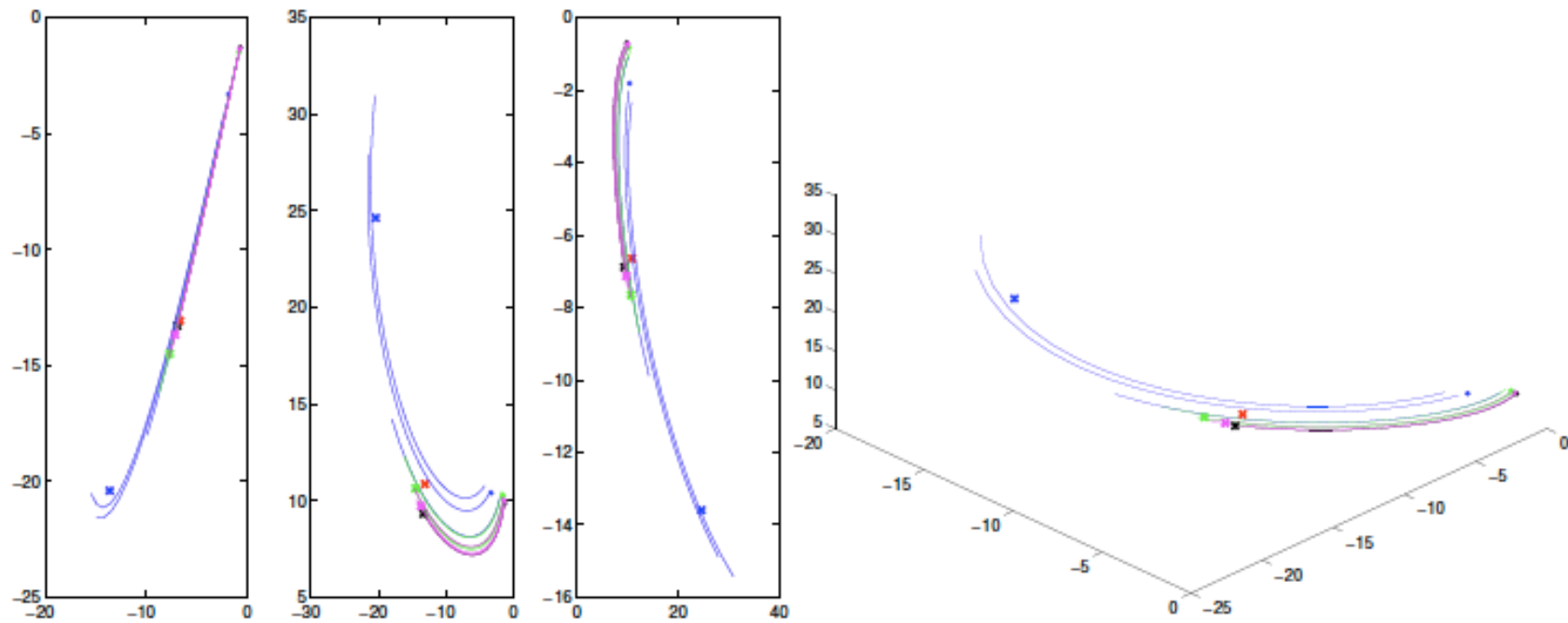
Example

Cycle 10 (2):



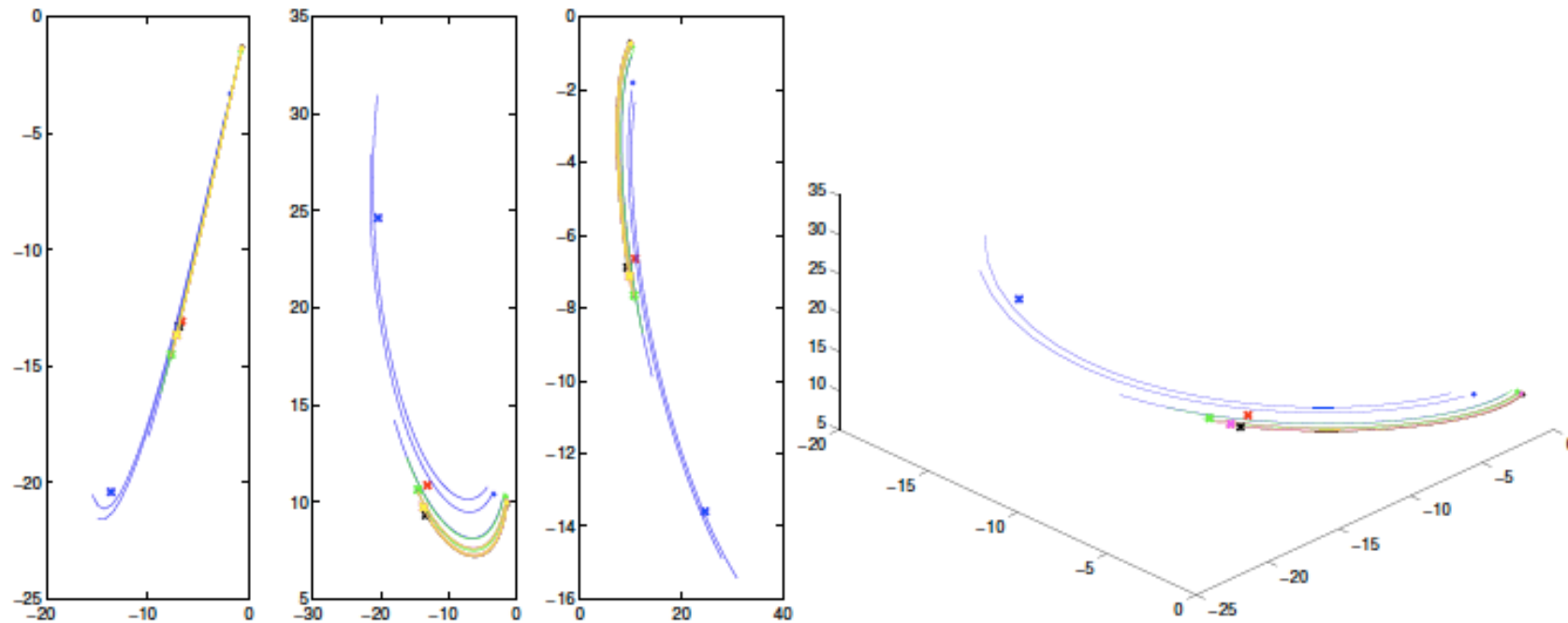
Example

Cycle 10 (3):



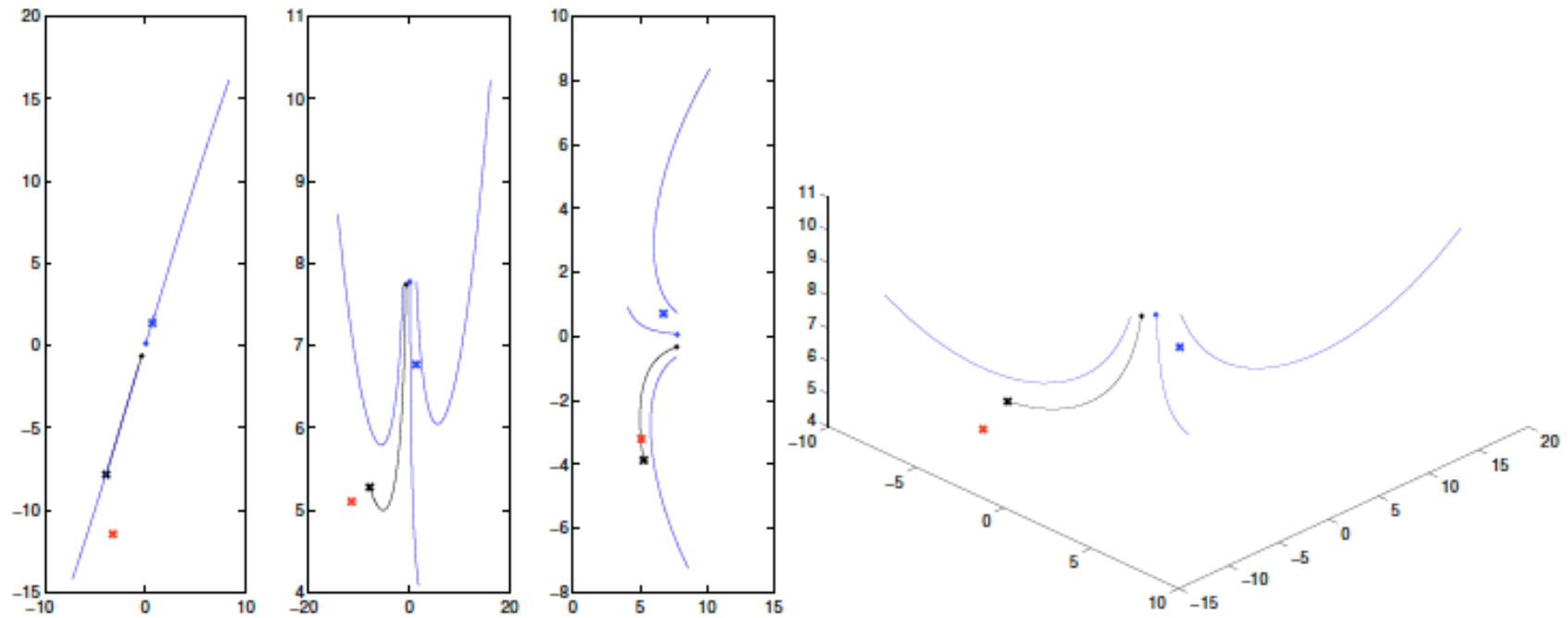
Example

Cycle 10 (4):



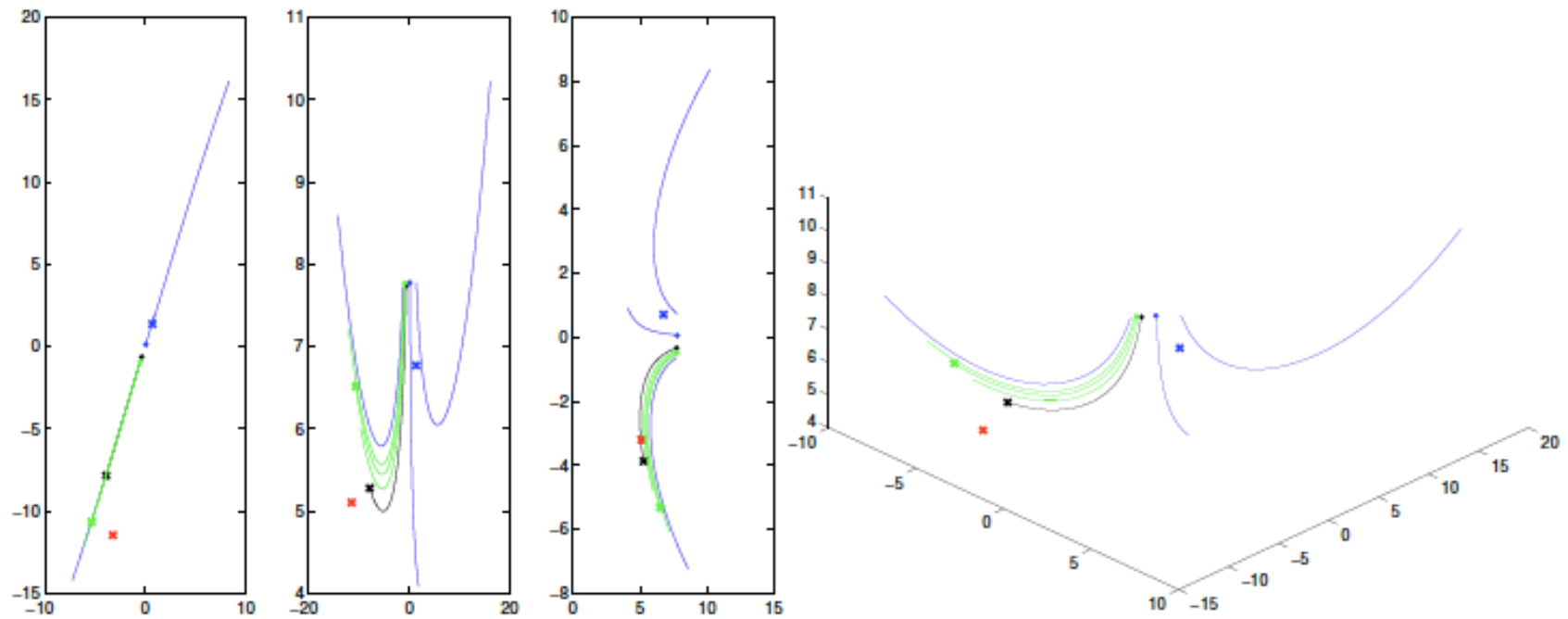
Example

Cycle 484 (1):



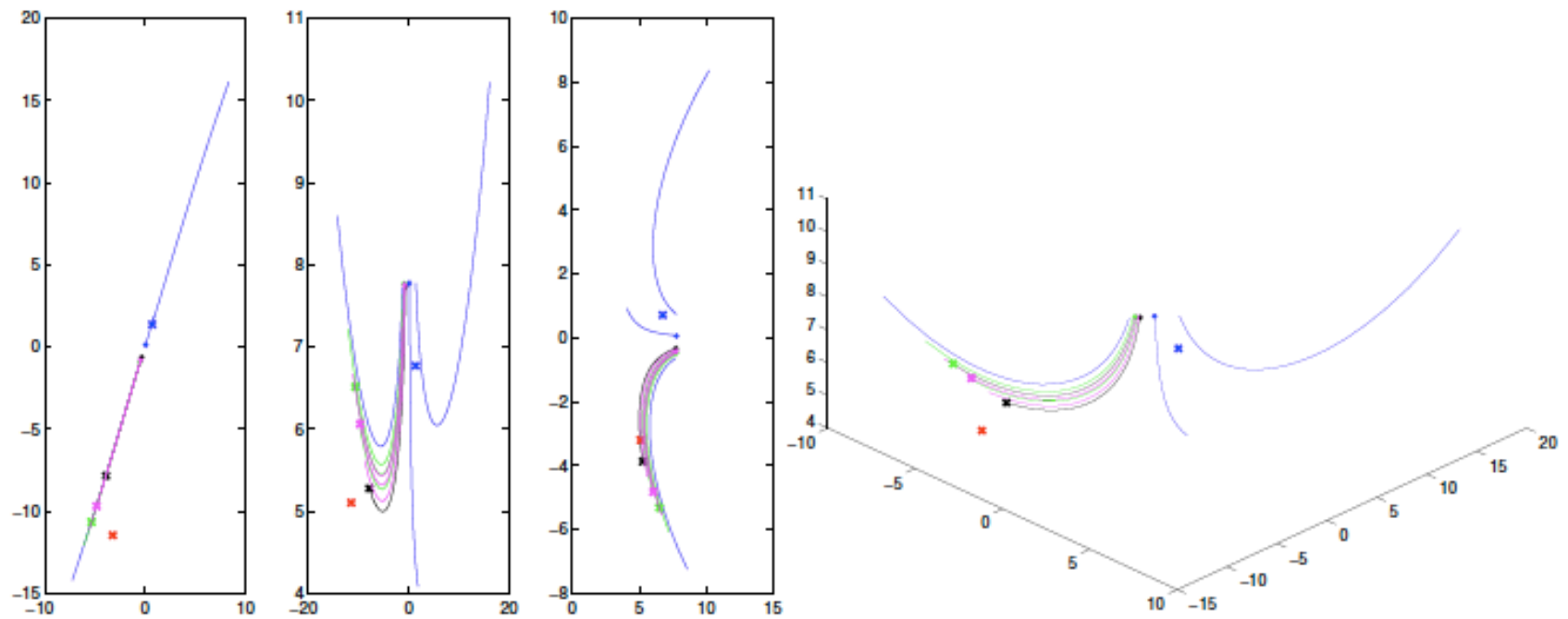
Example

Cycle 484 (2):



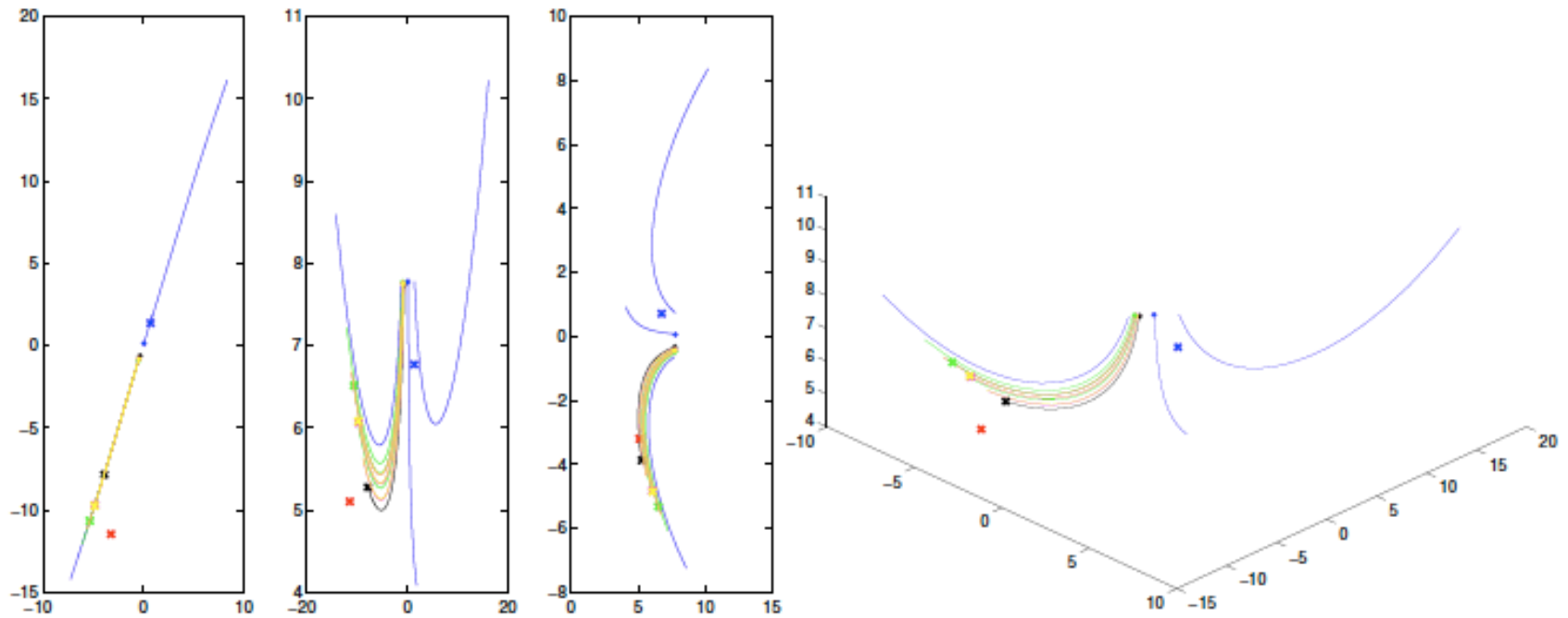
Example

Cycle 484 (3):



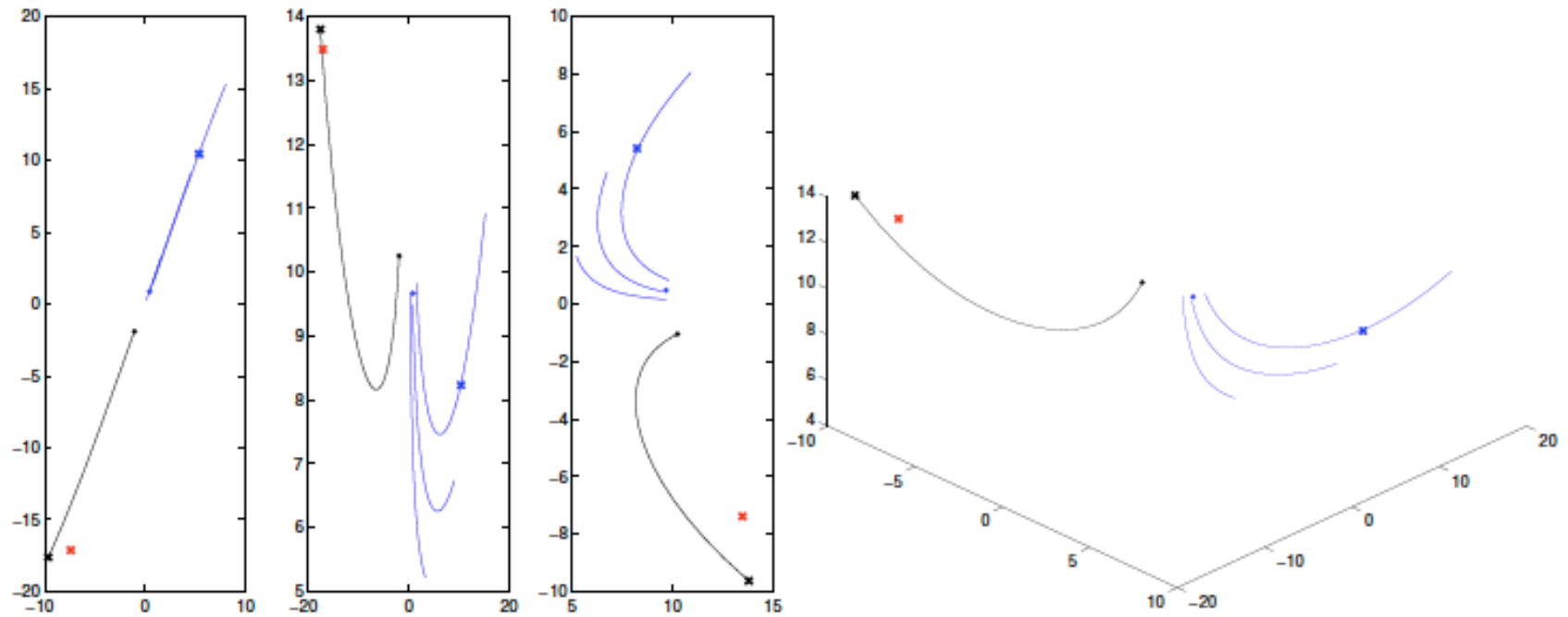
Example

Cycle 484 (4):



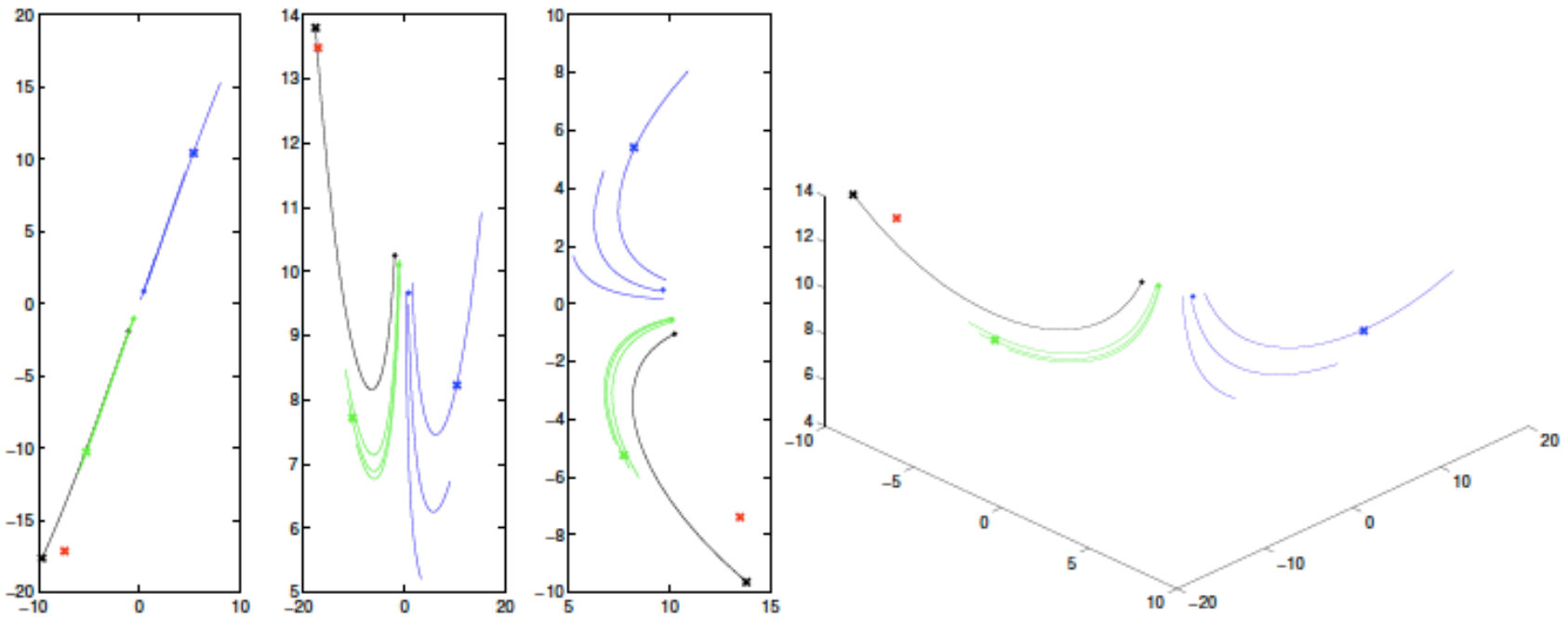
Example

Cycle 1392 (1):



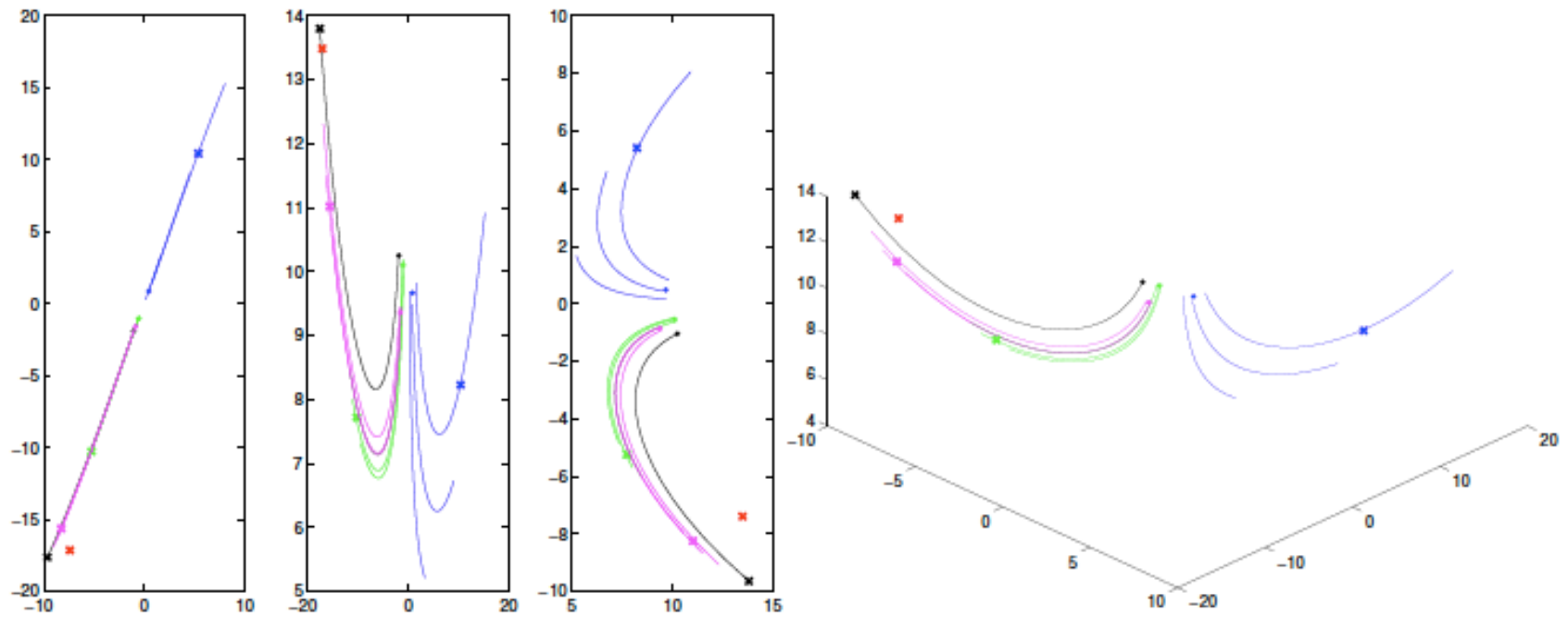
Example

Cycle 1392 (2):



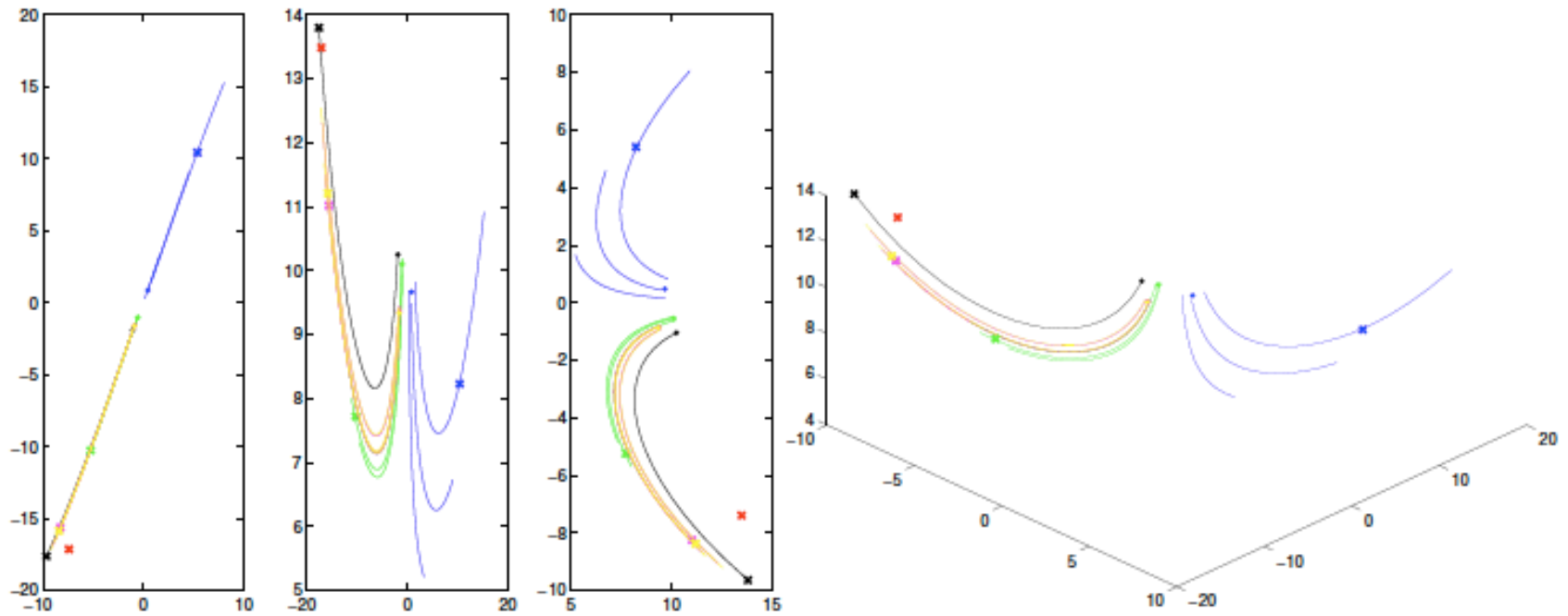
Example

Cycle 1392 (3):



Example

Cycle 1392 (4):



Experiment 1: L3, long window

(as in Miller et al. 1994)

scheme estimate	EnKF	IEnKF	IEKF
--------------------	------	-------	------

$m = 3$

analysis rmse	0.82	0.33	0.32
# iterations	1	2.8	2.7
inflation used	1.35	1.08	1.06

$m = 10$

analysis rmse	0.65	0.30	0.32
# iterations	1	2.6	2.7
inflation used	1.15	1.02	1.06

Kalnay et al. (2007):

- EnKF, $m = 3$: 0.71
- EnKF, $m = 6$: 0.59
- 4D-Var (assim. window = 75): 0.53

Yang et al. (2006):

- EKF: 0.63

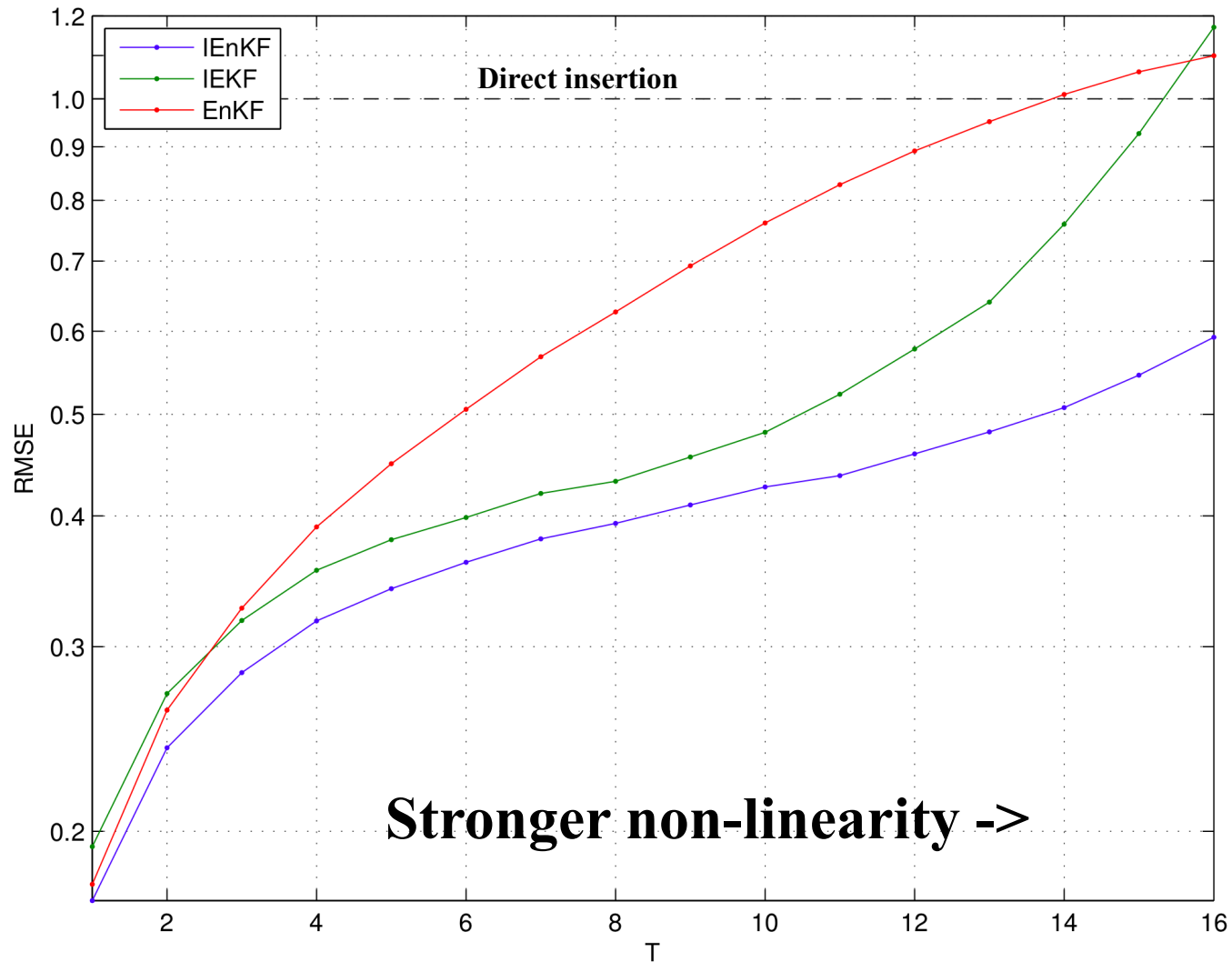
Kalnay and Yang,

<http://www.mpiyks-dresden.mpg.de/~ecodyc10/Contributions/Kalnay.pdf>:

- RIP, $m = ?$: 0.39

Benefits of Monte Carlo methods

RMS errors



■ Sakov, Oliver & Bertino, subm. MWR



Conclusions

- ▶ The iterative EnKF is a natural generalisation of the EnKF for strongly nonlinear systems
- ▶ In the linear case the IEnKF is equivalent to the EnKF (the second and following iterations return zero increments)
- ▶ The IEnKF outperforms other schemes in the experiments
- ▶ The IEnKF should be considered in the case of a well constrained system with nonlinear growth during the assimilation cycle (the propagation is initially nonlinear but becomes linear through iterations)
- ▶ The above implies a high degree of optimality (including a perfect or nearly perfect model)
- ▶ Similarly to the EnKF, the IEnKF can be easily transformed into the ensemble formulation of the IEKF
- ▶ Future work: include model error; localisation



Relevant tasks in MyOcean2

- Task 19.3.4 **Assimilation of new types satellite observations**
- (a) Assimilation of sea-ice properties (NERSC, Mercator, METNO, DMI). A strategy will be derived for combined assimilation of ice concentrations and thickness as from CryoSAT. Different anamorphic transformations will be tested (static, variable in time or space), the observation operator for sea-ice concentrations will be tested as a choice for the observations. Diagnostics to be checked: scatterplots, representers. The multivariate impact on thickness, surface temperature and salinity will be monitored.



Google image “Sangoma”



We have
nicer bugs!



Painless assimilation!