# A "Dressed" Ensemble Kalman Filter for Data Assimilation using Hybrid Coordinate Ocean Model in Pacific

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## **Outline**



- Background
- Dressed Ensemble Kalman Filter
- Experiments: model, data, schemes
- Results
- Summary

## **Background**

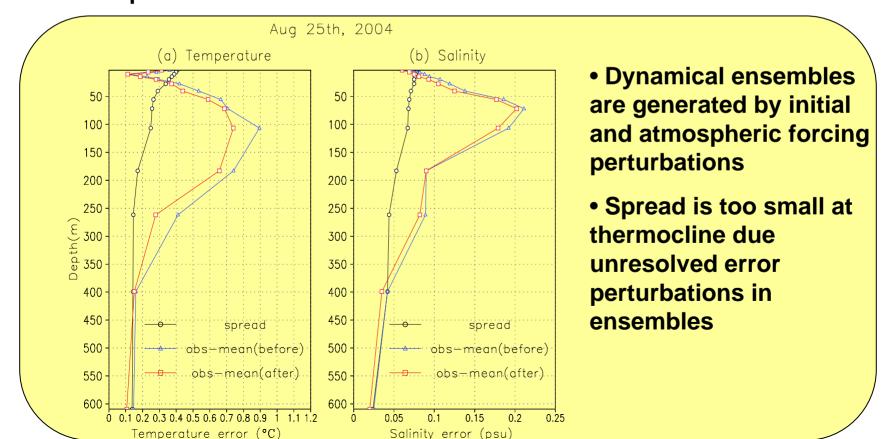


In EnKF, forecast error background covariance is defined by a set of dynamical ensemble

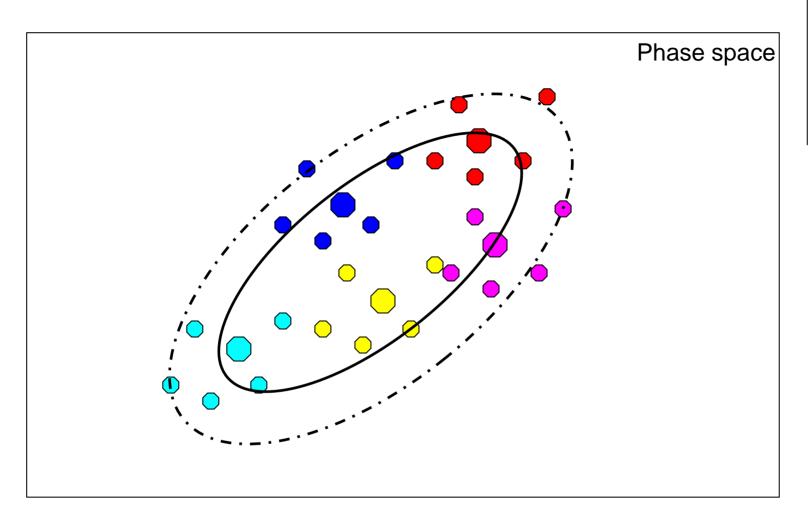
- + flow dependent;
- + nonlinearity of error evolution;
- + ...

#### **Problem**

- large computation cost even for "small" size ensemble (e.g., 100);
- error covariance generated confined into a subspace of small dimension.







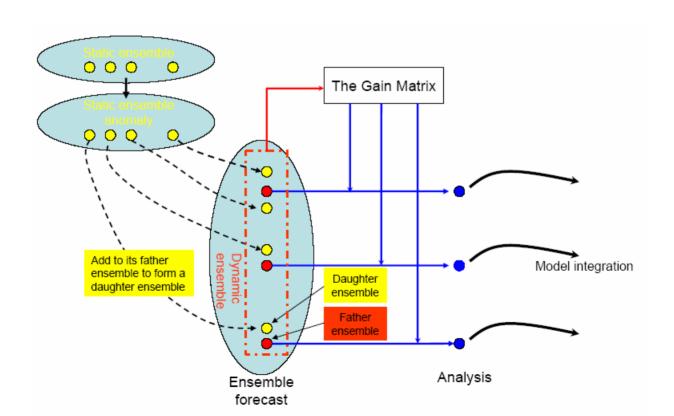


Idea of dressed dynamical ensemble with stationary ensemble by Roulston and Smith (2003).

## **Dressed EnKF**

Small size (e.g., 10) dynamical ensembles dressed by large size (e.g., 100) stationary ensembles.

- to reduce computation cost dramatically (1 vs 10);
- to compensate for unresolved error perturbations in dynamical ensembles.





# **Experiments**

Model HYCOM

**Domain** Pacific

ensemble)

H-Resolution 0.5°X 0.5°

Forcing ECMWF high frequency (6 hourly)

**COADS** climatology

**Assimilation** EnKF (dynamical ensemble size 100)

Schemes DrEnKF (10 dynamical ensemble, 100 stationary

ARGO Floats

Iongitude

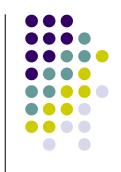
**Obs assimilated** Altimetry (CLS product, 1/3°, every 7 days)

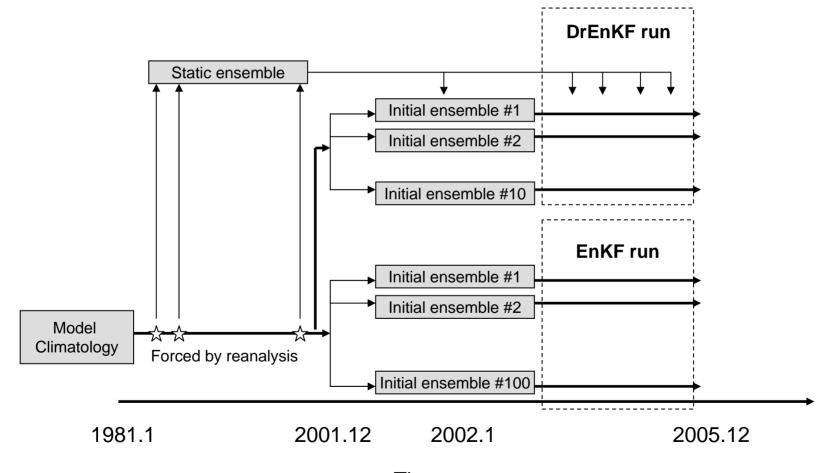
**Obs validation** ARGO T and S profiles, OISST

**Assimilation period** 2002.1-2005.12

EnKF: perturbations on initial state and atmospheric forcing;

DrEnKF: dynamical ensembles same as EnKF

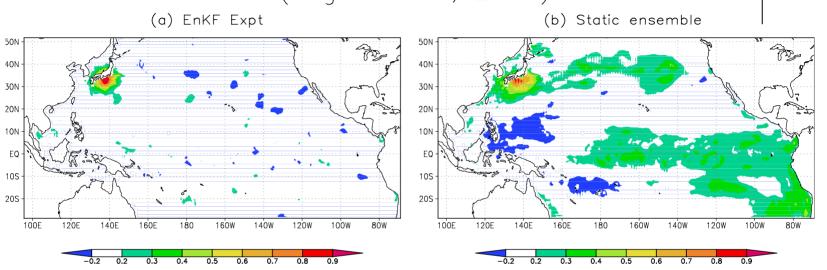


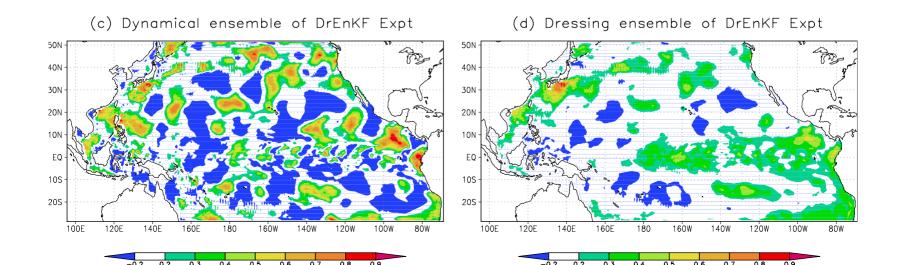


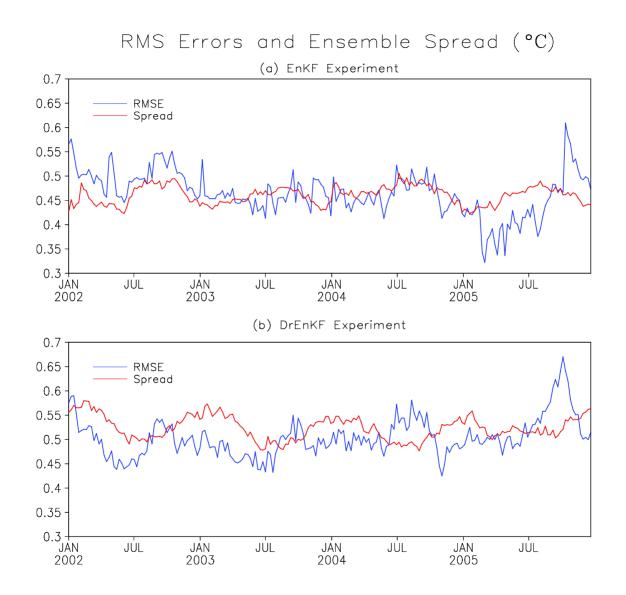
Time

### Results

SST Correlation with point (137E,33N) (August 18th, 2004)



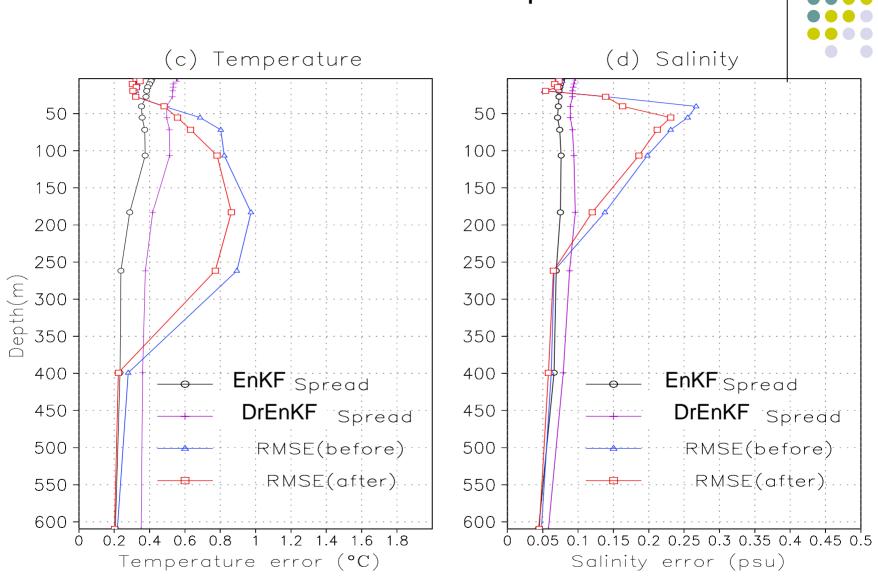


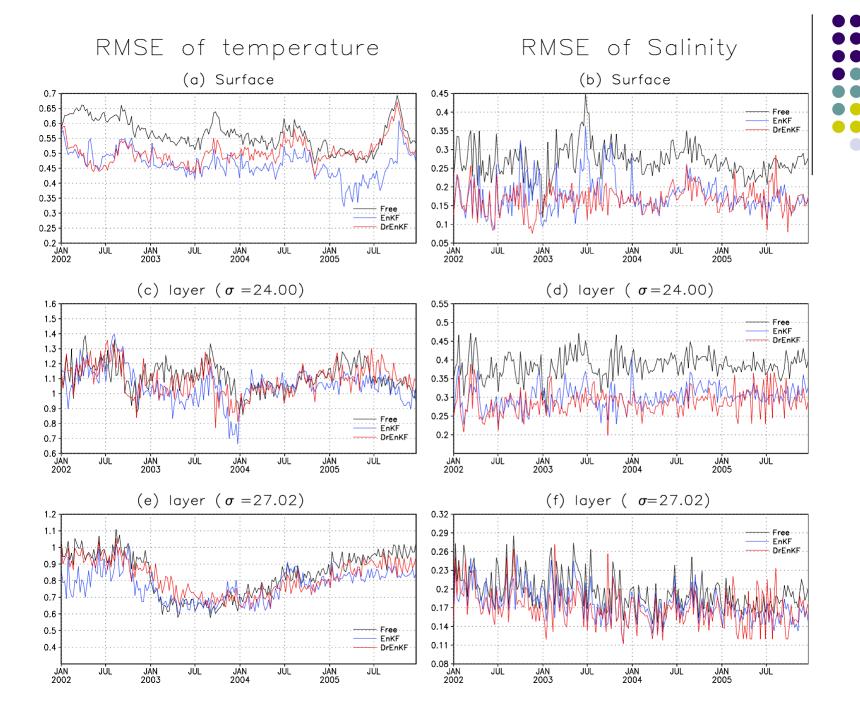


RMS errors against OISST.



#### RMSE of ensemble mean vs spread of DrEnKF





Layer	Free-running	EnKF		DrEnKF	
	T (°C)	T (°C)	Improvement percent (%)	T (°C)	Improvement percent (%)
1	0.564	0.484	14.2	0.506	10.3
2	0.982	0.930	5.3	0.935	4.8
3	0.987	0.936	5.2	0.936	5.2
4	1.001	0.946	5.5	0.948	5.3
5	1.011	0.967	4.4	0.969	4.2
6	1.046	1.009	3.5	1.008	3.6
7	1.124	1.088	3.2	1.100	2.2
8	1.172	1.141	2.7	1.161	0.9
9	1.173	1.170	0.3	1.172	0.1
10	1.215	1.205	0.8	1.209	0.5
11	1.194	1.181	1.1	1.182	1.0
12	1.419	1.412	0.5	1.416	0.2
13	1.341	1.271	5.2	1.287	4.0
14	0.725	0.674	7.0	0.694	4.3



	Free-running	EnKF		DrEnKF	
Layer	S (psu)	S (psu)	Improvement percent (%)	S (psu)	Improvement percent (%)
1	0.270	0.179	33.7	0.176	34.8
2	0.439	0.391	10.9	0.393	10.5
3	0.437	0.391	10.5	0.393	10.1
4	0.439	0.391	10.9	0.394	10.3
5	0.439	0.393	10.5	0.394	10.3
6	0.441	0.408	7.5	0.408	7.5
7	0.456	0.433	5.0	0.437	4.2
8	0.460	0.443	3.7	0.440	4.3
9	0.433	0.414	4.4	0.417	3.7
10	0.360	0.321	10.8	0.318	11.7
11	0.412	0.369	10.4	0.376	8.7
12	0.348	0.287	17.5	0.296	14.9
13	0.248	0.218	12.1	0.208	16.1
14	0.190	0.176	7.4	0.179	5.8

## **Summary**



- The idea of a "Dressed" ensemble is the combination of a dynamical ensemble with a small size (here 10 members) and a static ensemble.
- DrEnKF avoids the underestimation of errors as expected with a dynamical ensemble of small size.
- HYCOM data assimilation experiments show that the DrEnKF can dramatically reduce computation while maintain the performance.

# Thank you for your attention!



#### Parameters in assimilation experiments

Parameters	Description	Value	
N	Number of static ensemble members in DrEnKF Experimen	nt 100	• •
	Number of dynamical members in EnKF Experiment	100	1
M	Number of dynamical members in DrEnKF Experiment	10	
m	Maximum number of local observations	49	
$r_0$	Observations radius of influence	700 km	
$\sigma_{a}^{0}$	Observations error standard deviation	0.05 m	
$r_a^e$	Observations error decorrelation length	200 km	
$egin{array}{c} \sigma_{\!\!\!\!\!e} \ r_{\!\!\!\!\!e} \ r_h \end{array}$	Initial error Decorrelation length	1000 km	
$\sigma_{\!\!d}^{\!\!n}$	Layer thickness logarithmic standard deviation	10 %	
	Temperature logarithmic standard deviation in Mixed Layer	5 %	
$egin{array}{c} oldsymbol{\sigma}_T \ oldsymbol{r}_{vd} \end{array}$	Vertical correlation coefficient of layer thickness	2.0	
	Vertical correlation coefficient of temperature	3.0	
νT Υ	Decorrelation length of random forcing	10 grid cells	
$egin{array}{c} r_{_{\!$	Atmospheric temperature standard deviation	3 K	
$o_{\mathrm{T}}$	Atmospheric wind stress standard deviation	0.03 N/m2	
$o_{\Gamma}$	Wind speed standard deviation	1.6 m/s	

0.2 W/m2

3 days

Radiative flux standard deviation

Time correlation scale