Hybrid Variational-Ensemble Data Assimilation for Tropical Cyclone Forecasts: ongoing efforts and plans

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Outline

Introduction: what and why hybrid?

Results of hybrid DA for

a). GFS Hurricane Track forecast 2010: ENS4DVAR vs. ENS3DVAR vs. GSI b). Diagnostics study on why hybrid provided better track forecast for Ike 2008

c). High resolution radar data assimilation for hurricane Ike 2008

d). Integration of global hybrid GSI-EnKF with regional: an encouraging story

Plan for HWRF

Hybrid GSI-EnKF DA system



Why Hybrid? "Best of both worlds"

	VAR (3D, 4D)	EnKF	hybrid	References (examples)
Benefit from use of flow dependent ensemble covariance instead of static B		X	X	Hamill and Snyder 2000; Wang et al. 2007b,2008ab, 2009b; Zhang et al. 2009; Buehner et al. 2010ab; Wang 2011;
Robust for small ensemble			x	Wang et al. 2007b, 2009b; Buehner et al. 2010b
Better localization for integrated measure, e.g. satellite radiance; radar with attenuation			X	Campbell et al. 2010
Easiness to add various constraints	x		x	
Outer loops	x		x	
More use of various existing capability in VAR	х		x	

Summarized in Wang 2010, MWR

• How to incorporate ensemble in GSI?

• Extended control variable method (Wang 2010, MWR):

$$J(\mathbf{x}_{1}, \boldsymbol{\alpha}) = \beta_{1}J_{1} + \beta_{2}J_{e} + J_{o}$$

$$= \beta_{1}\frac{1}{2}\mathbf{x}_{1}^{T}\mathbf{B}^{-1}\mathbf{x}_{1}^{'} + \beta_{2}\frac{1}{2}\boldsymbol{\alpha}^{T}\mathbf{C}^{-1}\boldsymbol{\alpha} + \frac{1}{2}(\mathbf{y}^{o'} - \mathbf{H}\mathbf{x}^{'})^{T}\mathbf{R}^{-1}(\mathbf{y}^{o'} - \mathbf{H}\mathbf{x}^{'})$$
Extra term associated with extended control variable
$$\mathbf{x}' = \mathbf{x}_{1}' + \sum_{k=1}^{K}(\boldsymbol{\alpha}_{k} \circ \mathbf{x}_{k}^{e})$$
Extra increment associated with ensemble

B 3DVAR static covariance; **R** observation error covariance; *K* ensemble size; **C** correlation matrix for ensemble covariance localization; \mathbf{x}_k^e *k*th ensemble perturbation; $\mathbf{x}_1^{'}$ 3DVAR increment; $\mathbf{x}^{'}$ total (hybrid) increment; $\mathbf{y}^{o'}$ innovation vector; **H** linearized observation operator; β_1 weighting coefficient for static covariance; β_2 weighting coefficient for ensemble covariance; $\boldsymbol{\alpha}$ extended control variable.



- Like traditional 4DVAR, 4D analyses are obtained by fitting observations spanning the assimilation window.
- Unlike traditional 4DVAR, ens4dvar does not need to develop the tangent linear and adjoint of the forecast model (Tian et al. 2008, Liu et al. 2008).



RMSE of global forecasts by GFS



Significant improvement of ens3dvar hybrid and ens4dvar hybrid over GSI
 ens4dvar showed further improvement over ens3dvar especially for wind

Hurricane track forecasts by GFS

2010 Aug.-Sep.

2010 hurricane



- Improvement of TC track forecasts by ens3dvar hybrid than GSI and further improvement by ens4dvar hybrid.
- Balance constraint in GSI hurt TC forecast of hybrid. Lei, Wang et al. 2011

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Radar hybrid DA for hurricane





Wind increment

3DVAR with untuned static covariance (3DVARa)

Hybrid with full ensemble covariance (hybrid1)





Temperature increment



Wind and pot. temperature analyses



Track and intensity forecasts





Wind and Precipitation forecasts





Integration of the global hybrid to the regional



Model: WRF ARW,

Ensemble size: 40

Observations:

operational data except satellite radiances

May 8-16, 2010

Successful integration



- The global hybrid GSI-EnKF system is successfully integrated with the regional WRF ARW model for operational Rapid Refresh system.
- Experience gained will help conduct the same integration and test for HWRF.



Summary

Ensemble-4DVAR (no tangent linear and adjoint needed) was developed for GSI and tested for GFS. Ensemble-4DVAR further improved upon the ENS3DVAR hybrid for TC track forecasts. Balance constraint in GSI hurt TC forecasts using hybrid.

Plan on quasi real time demo of ens4dvar for the global system 2012 with ESRL. Apply ens4dvar for the regional.

 The hybrid was also implemented for high resolution radar data assimilation for TC forecast and showed improvement over 3DVAR.

The global hybrid system was successfully integrated with the regional WRF ARW model for operational RR system.



A proposal to HFIP

- Improving High-Resolution Tropical Cyclone Prediction Using a Unified GSI-based Hybrid Ensemble-Variational Data Assimilation System for HWRF" by Xuguang Wang (PI) and Ming Xue (co-PI), University of Oklahoma
- The extension, application and extensive testing of GSI-based hybrid DA for the HWRF modeling system, at high resolutions, and including both environmental and inner-core observations, is the primary goal of this proposal.
- □ Specific approaches, execution plan, milestones, deliverables, collaboration with EMC, AOML, ESRL, DTC were proposed.

Hybrid GSI-EnKF DA system



Wang et al. 2011



- □ Leverage and contribute to the efforts by ESRL and AOML to interface and test the global EnKF with HWRF, and to add inner core data assimilation capability.
- Leveraging the framework from EMC, further extend and add various new capabilities of GSI-ECV for HWRF (hydrometeors, ENS4DVAR, balance weak constraint etc). 9 km domain was planned to be bigger than EMC to have less frequent domain changes.
- □ Select a retrospective case with airborne Doppler radar, ground-based radar and other inner-core observations available (with EMC and AOML).
- Extensive testing, tuning of the outer HWRF domain nested within the global hybrid (prepare global hybrid with ESRL and EMC).
- □ For the 3 km domain DA, we will gradually increase its degree of sophistication before reaching its own fully cycled hybrid DA.



Plans (continued)

- □ Conduct comprehensive evaluations and verifications for this case.
- Repeat with other cases and select the best configuration for HWRF hybrid DA based on these results.
- □ Prepare and run the system in (quasi) real time (stream 2) for 2012 season.
- Conduct tests with more retrospective cases and use the evaluation metrics from NHC to conduct extensive evaluation of retrospective cases in preparation for stream 1.5 testing.
- □ Prepare the system for real time runs (stream 1.5) for the 2013 season.
- □ Make the system available to DTC and help DTC to conduct further testing.

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Why hybrid produced better track forecast?



Wang 2011, WAF



Initial analyses: 500mb height



- The subtropical high in the 3DVAR analysis extended more to the south in the southwest quadrant of Gulf of Mexico than HYBRID.
- Weaker and smaller IKE estimated by 3DVAR.

Analysis increment difference

