# Comparing Local Ensemble Transform Kalman Filter with 4D-Var in a Quasi-geostrophic model

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- Ensemble Kalman Filter (eg. LETKF) and 4D-Var are DA methods which can take into account the "flow-dependent errors".
- The implementation of LETKF and 4D-Var are very different:
  - LETKF: treat model as a black box, *local*
  - 4D-Var: model dependent, *global*
- Compare the performance of LETKF and 4D-Var

## **Experiment setup**



#### Quasi-Geostrophic Model

(Rotunno and Bao, 1996;Morss,1999)

- Channel model, periodic in x
- Horizontal: 64x33, Vertical: 7 levels
- Model variables

potential vorticity (q) arranged at interior 5 levels, potential temperature ( $\theta$ ) at top and bottom levels

- Experiment setup
  - 3% observation coverage (64 obs.) simulated rawinsonde (u,v,t) at all 7 levels, every 12hour
  - Analysis cycle: 12 hours
  - Initial condition, 3D-Var analysis solution

#### **Data assimilation schemes**

- 3D-Var (Morss, 1999)
  - B<sub>3D-Var</sub> has been optimized and is time-independent
  - Observation error covariance, R, is diagonal: uncorrelated between observations and between variables
  - Used as the benchmark
- Ensemble-based hybrid scheme (Corazza et al., 2002, Yang et al. 2006)
  - $B_{3D-Var}$  is augmented by the a set of bred vectors (the flow dependent errors)  $B_{HYBD}=(1-\alpha) B_{3D-Var}+\alpha EE^{T}$ .  $\alpha$  is the hybrid coefficient

 $[\mathbf{I} + ((1 - \alpha)\mathbf{B}_{3DVAR} + \alpha \mathbf{E}\mathbf{E}^{\mathrm{T}})\mathbf{H}^{\mathrm{T}}\mathbf{R}^{-1}\mathbf{H}](x_{a} - x_{b}) = [(1 - \alpha)\mathbf{B}_{3DVAR} + \alpha \mathbf{E}\mathbf{E}^{\mathrm{T}}]\mathbf{H}^{\mathrm{T}}\mathbf{R}^{-1}(y - \mathbf{H}x_{b})$ 

– Implemented in the 3D-Var framework

#### **Data assimilation schemes**

- LETKF (Hunt et al., 2006)
  - An efficient method to implement Ensemble Kalman Filter
    - Perform in a local volume (19x19x7)
    - Compute matrix inverse in the space spanned by ensemble (ensemble size =40)
    - A random perturbations (3% vectors amplitude) is added to the ensemble vectors
- 4D-Var
  - The adjoint model is generated by TAMC, but need to correct several subtle bugs related to boundary conditions
  - $B_0$  needs to be optimized.

 $\mathbf{B}=\mathbf{0.02}\times\mathbf{B}_{\mathrm{3DVAR}}$ 

# Ensemble-based hybrid scheme vs. Variational-based scheme



• The hybrid scheme performs better because of its ability to include the dynamically evolving errors

• By localizing the BVs,  $\alpha$  increases and the hybrid scheme perform much better

#### **RMS** analysis/forecast errors



The performance of LETKF is better than 4D-Var with 12-hour but worse than 4D-Var with 24-hour window

#### **Computational costs**

• Computational time with 1 CPU

	3D-Var	HYBD	4D-Var		LETKF		
			12HR	24HR	L=5	L=7	L=9
RMS error (x10 <sup>-2</sup> )	1.44	0.70	0.56	0.35	0.48	0.45	0.39
Time (hour)	0.5	5.0	8.0	14.0	5.5	8.3	10.0

LETKF can be computed in parallel

# Error variance vs. ensemble spread



## The structures of analysis increment

- 4D-Var analysis increments vs. singular vector(SV)
  - SV is defined with potential enstrophy norm with a chosen optimization time
  - Compared at initial/final time



- LETKF analysis increment vs. bred vector(BV)
  - At the analysis time

#### Structure of analysis increments



The initial analysis increments in 4D-Var are very different from the final increments, which are more similar to the analysis increments in LETKF

#### **Relative improvement in spectral coordinates**



#### Summary

From the perfect model experiments with an analysis cycle of 12-hour, we show that

- The ensemble spread from LETKF is able to reflect well the error covariance structure.
- LETKF has the performance in between the results of 4D-Var with 12-hour and 24-hour window. 4D-Var has an advantage with a long window.
- The analysis increment from LETKF is very similar to the analysis increment of 4D-Var at the end of the assimilation window. Both strongly resemble the BV and final SV.
- Both LETKF and 4D-Var successfully improve the 3D-Var analysis in all scales. The improvement of LETKF of large scale is as good as the 4D-Var with 24-hour window.