On the selection of localization radius in ensemble filtering for multi-scale quasi-geostrophic dynamics

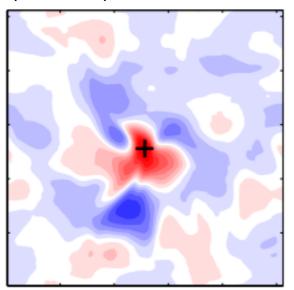
Yue (Michael) Ying, Fuqing Zhang, PennState/ADAPT

Jeffery Anderson, NCAR/DAReS

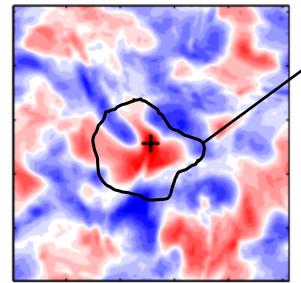
Group meeting, May 1 2017

In pursuit of an optimal localization:

"True" correlation (N=2000)



sample-estimated correlation (N=20)



localized analysis increment

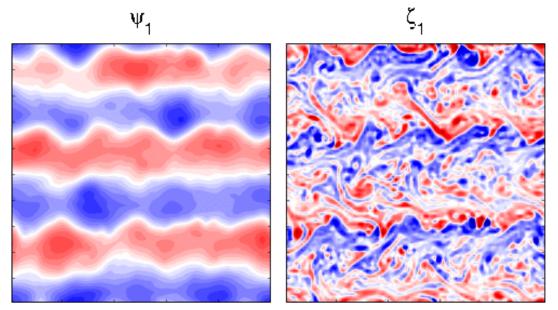
adaptive algorithms: Zhen and Zhang 2014 Perianez et al. 2014 Kirchgessner et al. 2014 Flowerdew 2015

localization radius depends on:

- ensemble size
- observing network properties (density, frequency, accuracy)
- underlying covariance structure, flow-dependency (Anderson 2007)

What is their relative importance for multi-scale dynamics?

Two-layer Quasi-Geostrophic Model



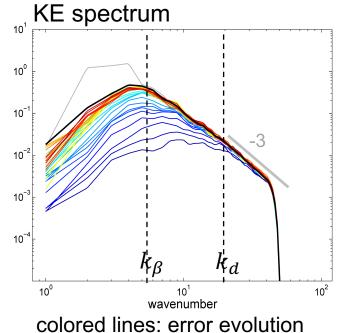
governing equations:

$$\frac{\partial q_1}{\partial t} + J(\psi_1, q_1) + U \frac{\partial q_1}{\partial x} + (\beta + k_d^2 U) \frac{\partial \psi_1}{\partial x} = 0,$$

$$\frac{\partial q_2}{\partial t} + J(\psi_2, q_2) - U \frac{\partial q_2}{\partial x} + (\beta - k_d^2 U) \frac{\partial \psi_2}{\partial x} + r \nabla^2 \psi_2 = 0,$$

$$q_1 = \nabla^2 \psi_1 + \frac{k_d^2}{2} (\psi_2 - \psi_1),$$

$$q_2 = \nabla^2 \psi_2 - \frac{k_d^2}{2}(\psi_2 - \psi_1),$$

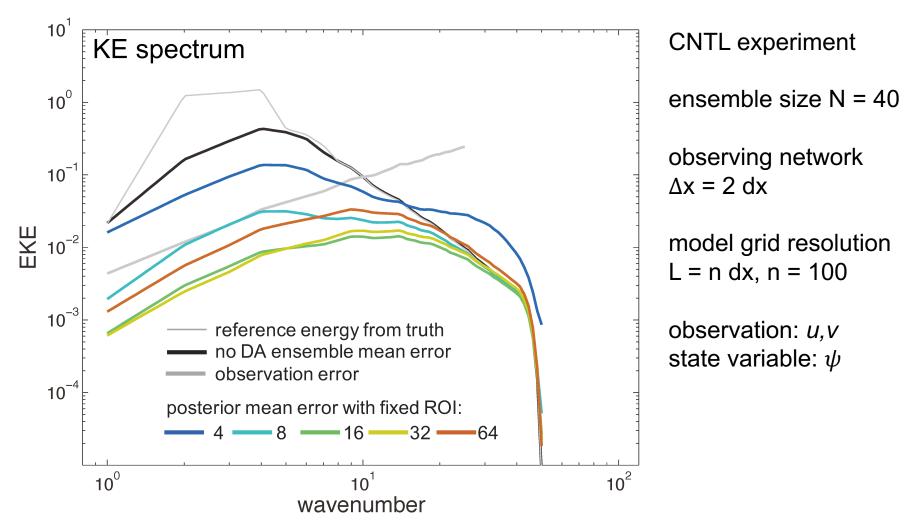


colored lines: error evolution gray: reference energy

capture the essence of large-scale atmospheric dynamics

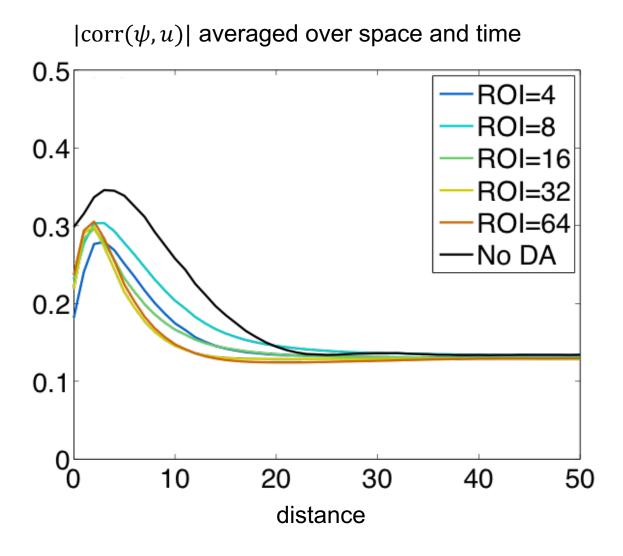
simple enough for sensitivity experiments

Finding best radius of influence (ROI) by trial-and-error



- best ROI is 16
- larger ROI is suboptimal for all scales
- smallest ROI introduce noise in the unobserved scales

Mean absolute correlation (MAC) as a function of distance

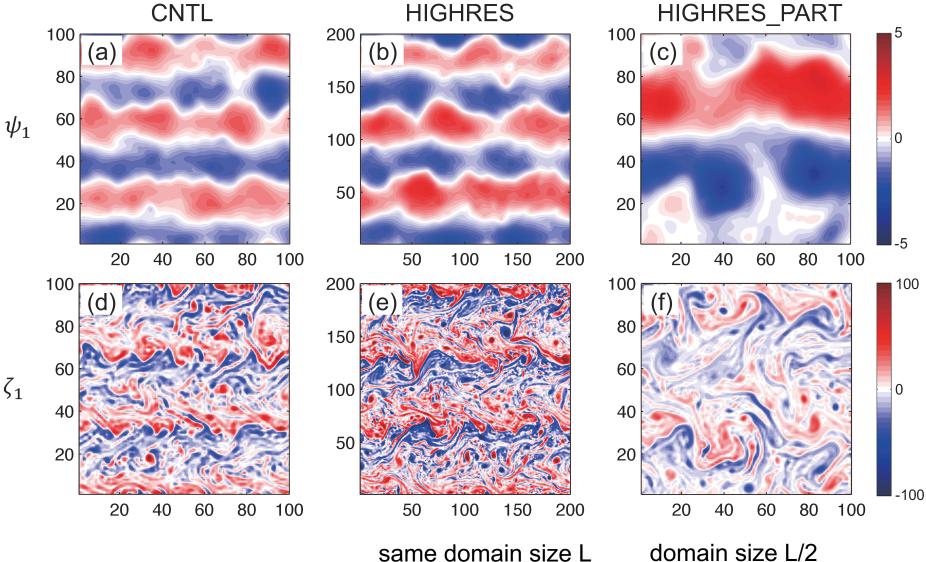


No DA correlation scale remains large

Asymptotic value of MAC is determined by ensemble size N.

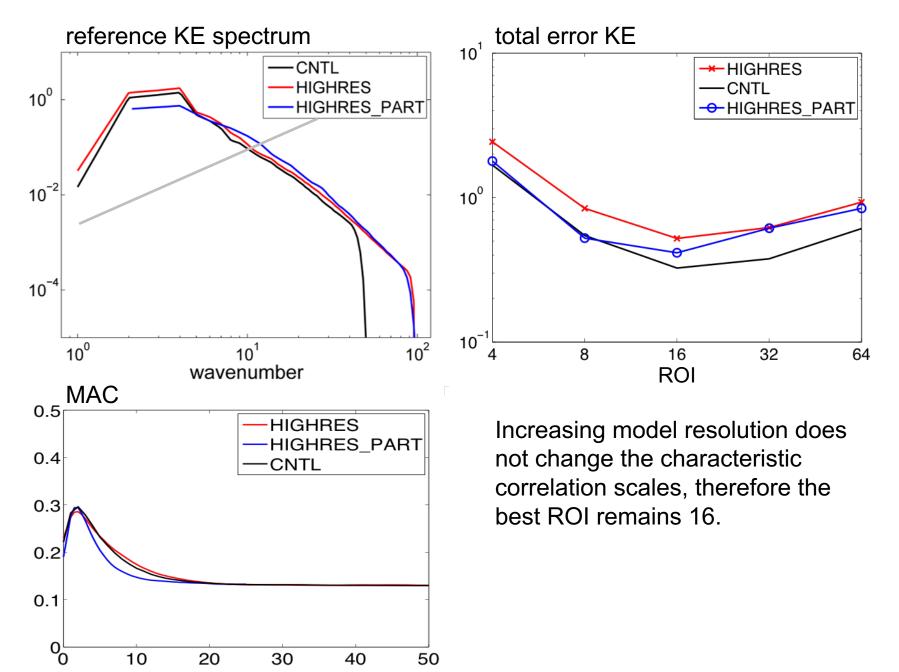
For the best ROI = 16, at distance = 16 the MAC is close enough to the asymptotic value.

Sensitivity to model resolution

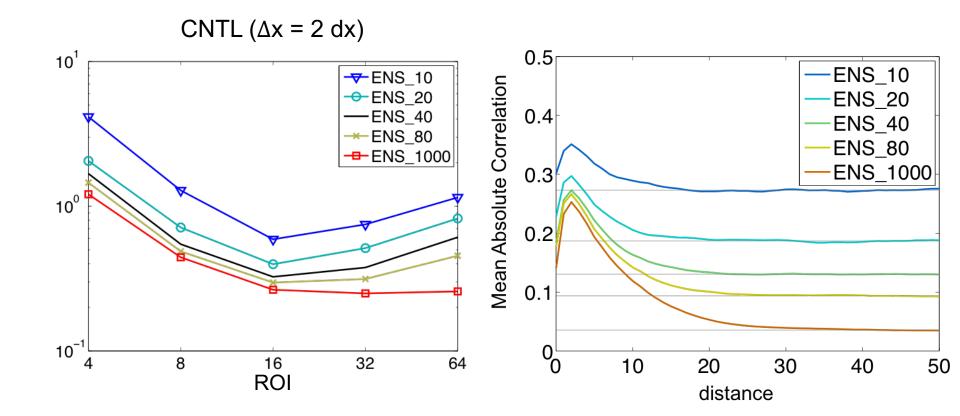


same domain size Ldomain size L/2grid spacing dx/2grid spacing dx/2

Sensitivity to model resolution



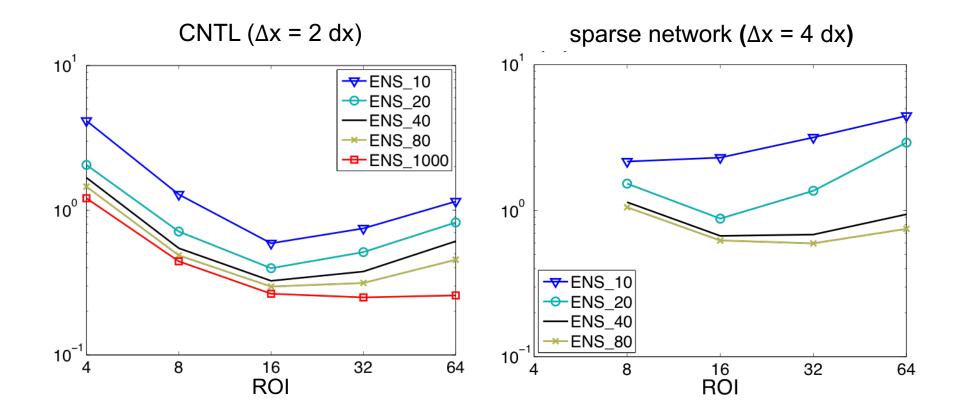
Sensitivity to ensemble size



sampling error decreases as ensemble size increase

the favorable range of ROIs is wider for larger ensemble size

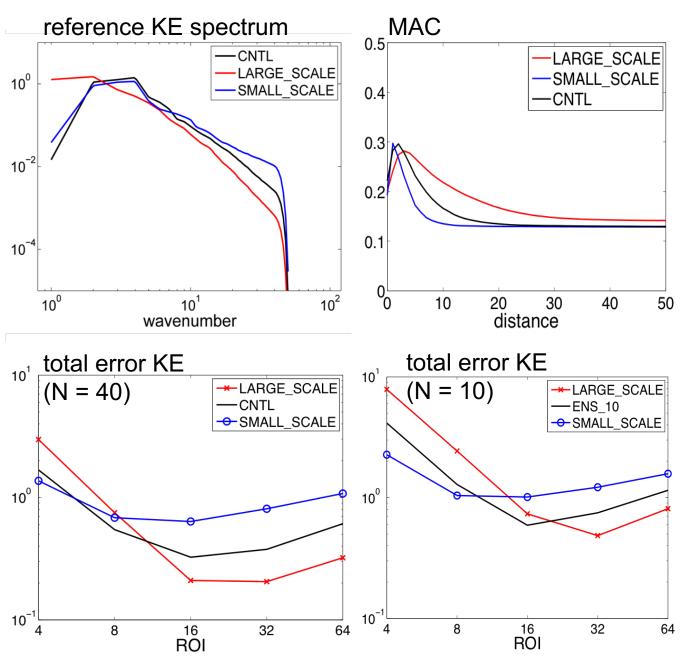
Sensitivity to ensemble size



shorter ROI should be used for smaller ensemble size (Anderson 2007 and many others)

For denser observing network (left), the small ensemble size seems to have more tolerance to sampling error.

Sensitivity to characteristic correlation scales



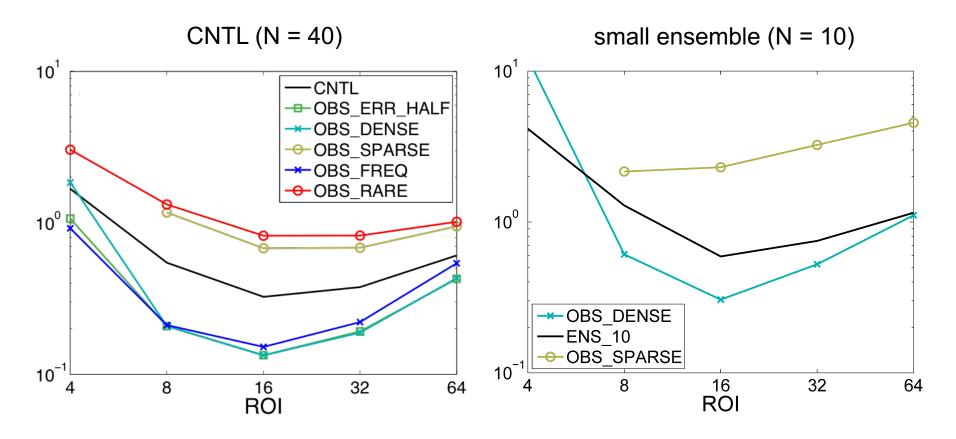
Changing characteristic correlation scale by shifting spectral peak,

more energy in small scale => shorter correlation scale.

The best ROI shifts to larger value if correlation scale is larger.

This effect is more evident for smaller ensemble

Sensitivity to observing network



Contradicting results:

- Denser network requires shorter ROIs (Fuqing's experience, Dong et al. 2011, Kirchgessner et al. 2014, Perianez et al. 2014)
- Denser network has broader optimal localization function (Anderson 2007)

Similar effect of increasing observation density (OBS_DENSE), time frequency (OBS_FREQ), or reducing observation error (OBS_ERR_HALF).

Concluding remarks

- A comprehensive sensitivity experiment is conducted using QG model to investigate how best-performing localization radius changes with:
- model resolution (not sensitive)

- ensemble size (small N requires shorter ROI, especially for sparse network)

 observing network
(sparser network requires shorter ROI if N is small; but can use longer ROI if N is large enough?)

- characteristic correlation scale (determined by dynamics) (an important factor, especially for small N)