



# Fast Implementation of Localization for Ensemble DA

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# EnKF

The formula of EnKF without localization

$$\mathbf{x}' = \mathbf{b} \mathbf{r}^T (\mathbf{O} + \mathbf{r} \mathbf{r}^T)^{-1} \mathbf{y}'_{obs}$$

**Localization:**

$$\mathbf{x}' = \boldsymbol{\rho}_{xy} \circ (\mathbf{b} \mathbf{r}^T) (\mathbf{O} + \boldsymbol{\rho}_{yy} \circ (\mathbf{r} \mathbf{r}^T))^{-1} \mathbf{y}'_{obs}$$

where

$\boldsymbol{\rho}_{xy}$ ,  $\boldsymbol{\rho}_{yy}$  : Filtering matrices;

$\circ$  : Schür product operator

$$\left\{ \begin{array}{l} \mathbf{b} = \frac{1}{\sqrt{n-1}} (\mathbf{x}_1 - \bar{\mathbf{x}}, \mathbf{x}_2 - \bar{\mathbf{x}}, \dots, \mathbf{x}_n - \bar{\mathbf{x}}), \\ \mathbf{r} = \frac{1}{\sqrt{n-1}} (\mathbf{y}_1 - \bar{\mathbf{y}}, \mathbf{y}_2 - \bar{\mathbf{y}}, \dots, \mathbf{y}_n - \bar{\mathbf{y}}), \end{array} \right. \quad \begin{array}{l} \bar{\mathbf{x}} = \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i \\ \bar{\mathbf{y}} = \frac{1}{n} \sum_{i=1}^n \mathbf{y}_i \end{array}$$

**The elements of filter matrix  $\rho_{xy}$  (similar for  $\rho_{yy}$ ):**

$$\rho_{i,j} = C_0(d_{i,j}^h / d_0^h) \cdot C_0(d_{i,j}^v / d_0^v)$$

$$(i = 1, 2, \dots, m_y; j = 1, 2, \dots, m_x) \quad ,$$

**where the filtering function  $C_0$  is defined as (Gaspari and Cohn, 1999)**

$$C_0(r) = \begin{cases} -\frac{1}{4}r^5 + \frac{1}{2}r^4 + \frac{5}{8}r^3 - \frac{5}{3}r^2 + 1, & 0 \leq r \leq 1 \\ \frac{1}{12}r^5 - \frac{1}{2}r^4 + \frac{5}{8}r^3 + \frac{5}{3}r^2 - 5r + 4 - \frac{2}{3}r^{-1}, & 1 < r \leq 2, \\ 0, & 2 < r \end{cases}$$

**$d_{i,j}^h$  and  $d_{i,j}^v$  respectively represent the horizontal and vertical distances between the  $i$ -th and  $j$ -th row vectors;  $d_0^h$  and  $d_0^v$  are the horizontal and vertical Schür radius.**

# Issues caused by Localization

Before localization, it is easy to get the solution:

$$\mathbf{x}' = \mathbf{b} \mathbf{r}^T (\mathbf{O} + \mathbf{R})^{-1} \mathbf{y}'_{obs} = \mathbf{b} \mathbf{r}^T (\mathbf{O} + \mathbf{r} \mathbf{r}^T)^{-1} \mathbf{y}'_{obs}$$

After localization, huge computing time is required for the solution:

$$m_x \times n \quad (n \sim 10^2)$$

$$m_y \sim 10^5$$

$$m_x \sim 10^7$$

$$\mathbf{x}' = \boldsymbol{\rho}_{xy} \circ (\mathbf{b} \mathbf{r}^T) (\mathbf{O} + \boldsymbol{\rho}_{yy} \circ (\mathbf{r} \mathbf{r}^T))^{-1} \mathbf{y}'_{obs}$$

$$m_x \times m_y$$

$$m_y \times n$$

diagonal

$$m_y \times m_y$$

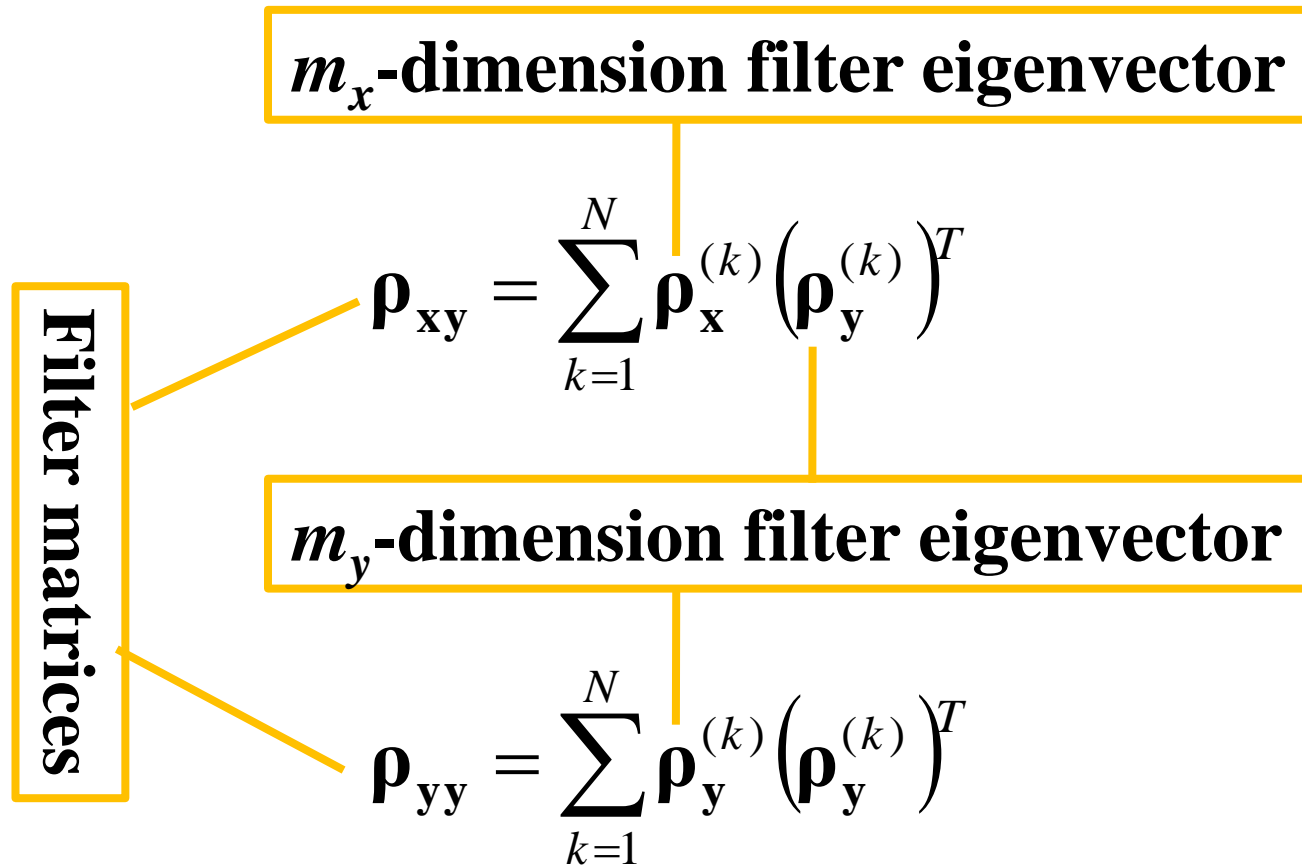
# The extra multiplications due to localization

$$m_x \times m_y \times n + m_y \times m_y \times n \times l \sim 10^{14}$$

## Available solution to the issue

- 1) To split the observation vector into a combination of low-dimension sub-vectors
- 2) To sequentially assimilate the low-dimension observation sub-vectors

# Our new method is to split filter matrix in localization



$N$  is the number of the major modes of the filter matrix

# Expansion of filter function in localization

Filter function:

Base function

$$C(x_1, x_2) = \sum_{n=1}^{\infty} a_n e_n(x_1) e_n(x_2)$$

$$\int_0^L w(x) e_n(x) e_m(x) dx = \begin{cases} 1 & \text{if } n = m \\ 0 & \text{if } n \neq m \end{cases}$$

Orthogonal

Weighting function

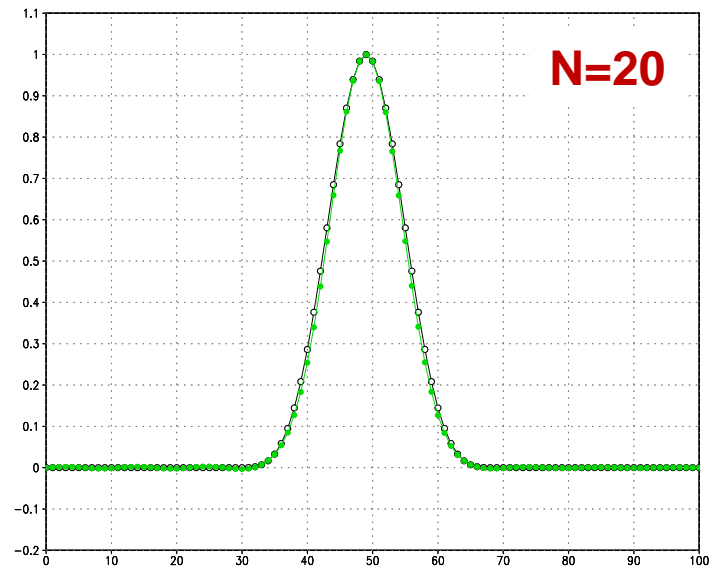
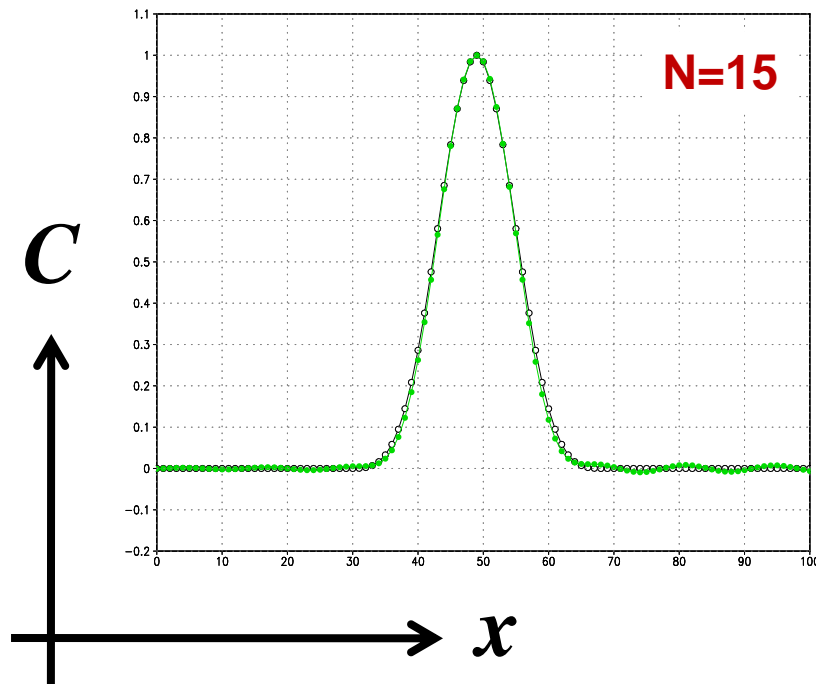
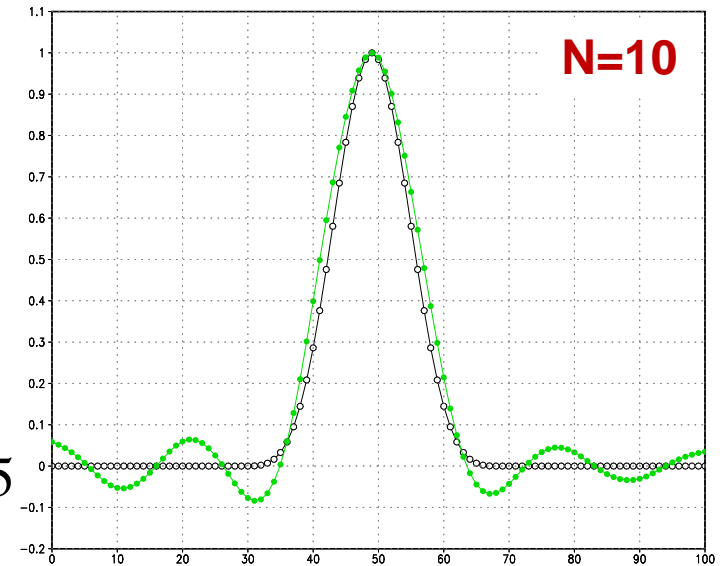
$$a_n = \int_0^L \int_0^L w(x_1) w(x_2) C(x_1, x_2) e_n(x_1) e_n(x_2) dx_1 dx_2$$

# One-Dimension Filter

**Green:**  $C_N(x, x_0) = \sum_{n=1}^N a_n e_n(x) e_n(x_0)$

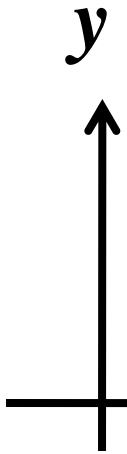
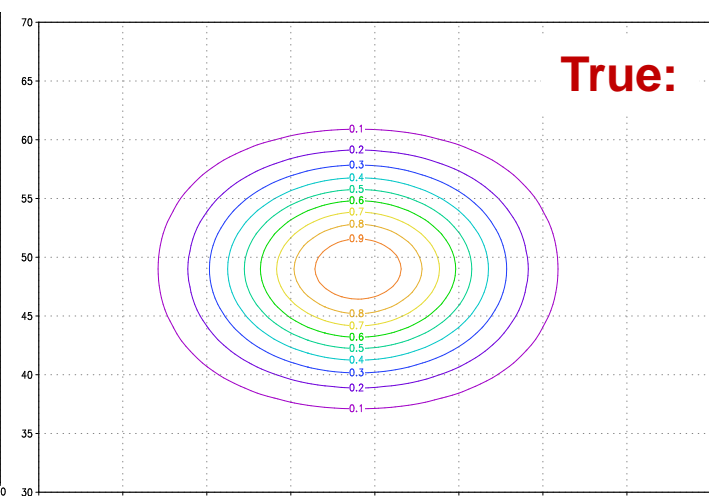
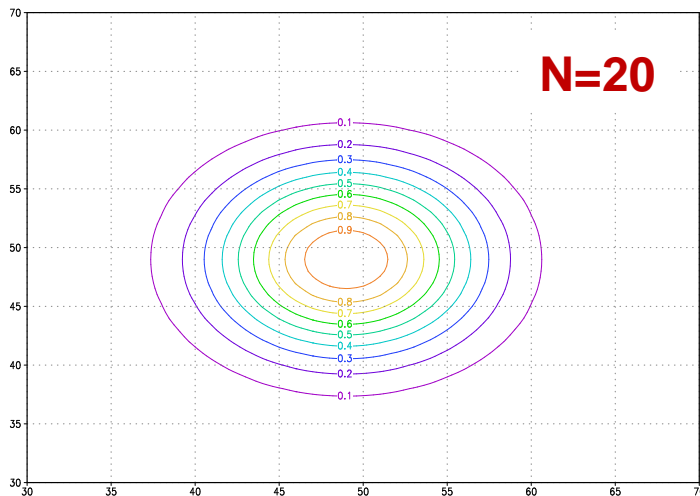
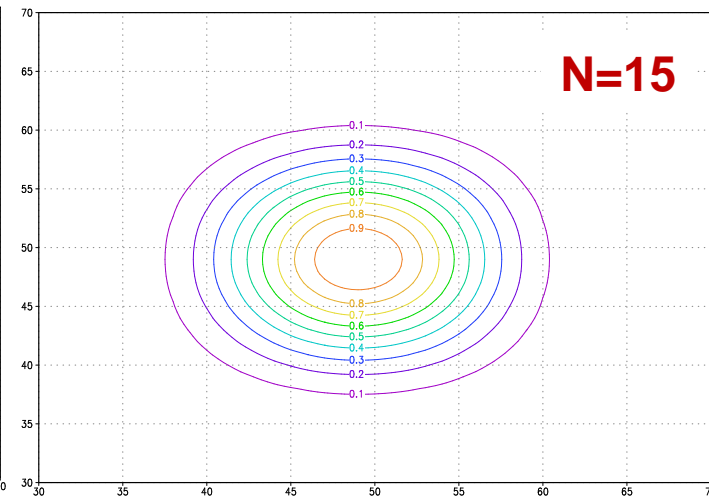
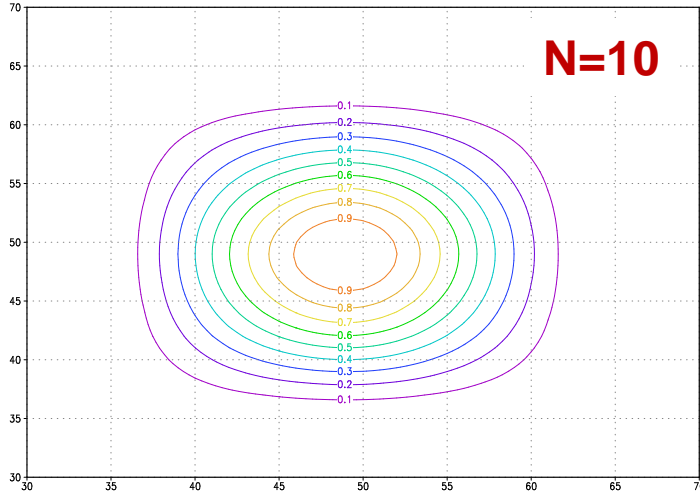
**Black:**  $C_0(r)$

**where**  $r = |x - x_0| / d_0$ ,  $x_0 = 49$ ,  $d_0 = 5$





# Two-Dimension Filter



$$C_N(x, x_0; y, y_0) = \sum_{n=1}^N \sum_{m=1}^N a_{n,m} e_n(x) e_n(x_0) e_m(y) e_m(y_0)$$

$$r = \sqrt{(x - x_0)^2 + (y - y_0)^2} / d_0, \quad x_0 = 49, \quad y_0 = 49, \quad d_0 = 5$$

# EnKF with extended samples

$$\mathbf{x}' = \tilde{\mathbf{b}} (\mathbf{I}_{\tilde{n} \times \tilde{n}} + \tilde{\mathbf{r}}^T \mathbf{O}^{-1} \tilde{\mathbf{r}})^{-1} \tilde{\mathbf{r}}^T \mathbf{O}^{-1} \mathbf{y}'_{obs}$$

where

$$\mathbf{b} = \frac{1}{\sqrt{\tilde{n} - 1}} (\boldsymbol{\rho}_{\mathbf{x}}^{(1)} \circ \mathbf{x}_1 - \bar{\mathbf{x}}, \boldsymbol{\rho}_{\mathbf{x}}^{(1)} \circ \mathbf{x}_2 - \bar{\mathbf{x}}, \dots, \boldsymbol{\rho}_{\mathbf{x}}^{(1)} \circ \mathbf{x}_n - \bar{\mathbf{x}}, \\ \boldsymbol{\rho}_{\mathbf{x}}^{(2)} \circ \mathbf{x}_1 - \bar{\mathbf{x}}, \boldsymbol{\rho}_{\mathbf{x}}^{(2)} \circ \mathbf{x}_2 - \bar{\mathbf{x}}, \dots, \boldsymbol{\rho}_{\mathbf{x}}^{(2)} \circ \mathbf{x}_n - \bar{\mathbf{x}}, \\ \vdots \\ \boldsymbol{\rho}_{\mathbf{x}}^{(N)} \circ \mathbf{x}_1 - \bar{\mathbf{x}}, \boldsymbol{\rho}_{\mathbf{x}}^{(N)} \circ \mathbf{x}_2 - \bar{\mathbf{x}}, \dots, \boldsymbol{\rho}_{\mathbf{x}}^{(N)} \circ \mathbf{x}_n - \bar{\mathbf{x}})$$

$$\bar{\mathbf{x}} = \frac{1}{\tilde{n}} \sum_{k=1}^N \sum_{i=1}^n \boldsymbol{\rho}_{\mathbf{x}}^{(k)} \circ \mathbf{x}_i \quad (\tilde{n} = n \times N)$$

$$\begin{aligned}
\mathbf{r} = \frac{1}{\sqrt{\tilde{n} - 1}} & (\boldsymbol{\rho}_{\mathbf{y}}^{(1)} \circ \mathbf{y}_1 - \bar{\mathbf{y}}, \boldsymbol{\rho}_{\mathbf{y}}^{(1)} \circ \mathbf{y}_2 - \bar{\mathbf{y}}, \dots, \boldsymbol{\rho}_{\mathbf{y}}^{(1)} \circ \mathbf{y}_n - \bar{\mathbf{y}}, \\
& \boldsymbol{\rho}_{\mathbf{y}}^{(2)} \circ \mathbf{y}_1 - \bar{\mathbf{y}}, \boldsymbol{\rho}_{\mathbf{y}}^{(2)} \circ \mathbf{y}_2 - \bar{\mathbf{y}}, \dots, \boldsymbol{\rho}_{\mathbf{y}}^{(2)} \circ \mathbf{y}_n - \bar{\mathbf{y}}, \\
& \vdots \\
& \boldsymbol{\rho}_{\mathbf{y}}^{(N)} \circ \mathbf{y}_1 - \bar{\mathbf{y}}, \boldsymbol{\rho}_{\mathbf{y}}^{(N)} \circ \mathbf{y}_2 - \bar{\mathbf{y}}, \dots, \boldsymbol{\rho}_{\mathbf{y}}^{(N)} \circ \mathbf{y}_n - \bar{\mathbf{y}})
\end{aligned}$$

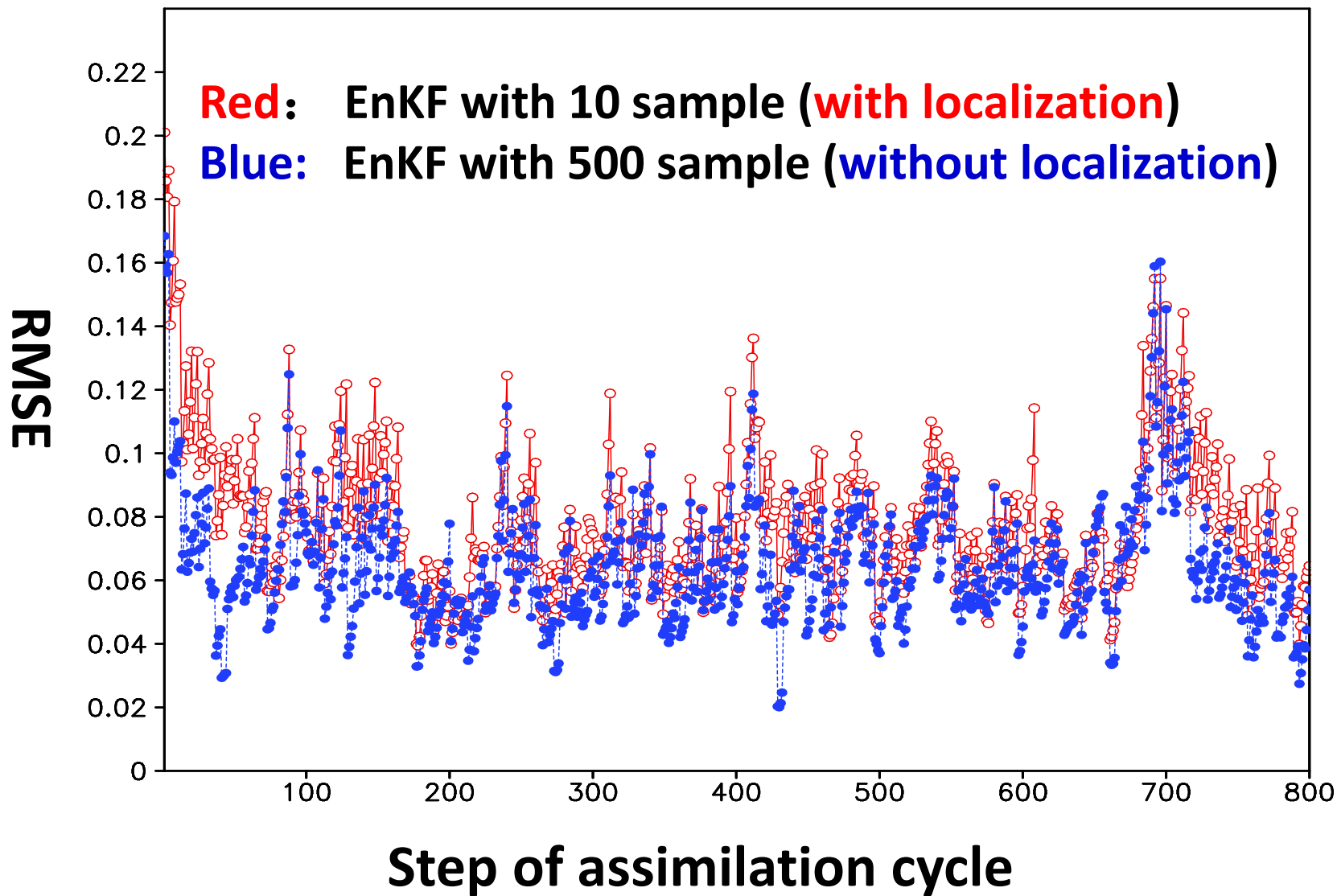
$$\bar{\mathbf{y}} = \frac{1}{\tilde{n}} \sum_{k=1}^N \sum_{i=1}^n \boldsymbol{\rho}_{\mathbf{y}}^{(k)} \circ \mathbf{y}_i \quad (\tilde{n} = n \times N)$$

# Experiment with simple model

## Lorenz-96 model:

$$\frac{dX_j}{dt} = (X_{j+1} - X_{j-2})X_{j-1} - X_j + F$$

- ◆ **Spatial coordinate:**  $j = 1, \dots, M$  ( $M = 40$ )
- ◆ **Forcing parameter:**  $F = 8$
- ◆ **Solution scheme:** 4<sup>th</sup> -order Runge-Kutta scheme
- ◆ **Time stepsize:** 0.05 time unit
- ◆ **Periodic boundary conditions:**  $X_{j+40} = X_j$
- ◆ **Note:** 0.05 time unit  $\sim$  6h (Lorenz and Emanuel, 1998)



# Experiment based on an operational prediction model: AREM

**CTRL:** No assimilation with prediction as IC

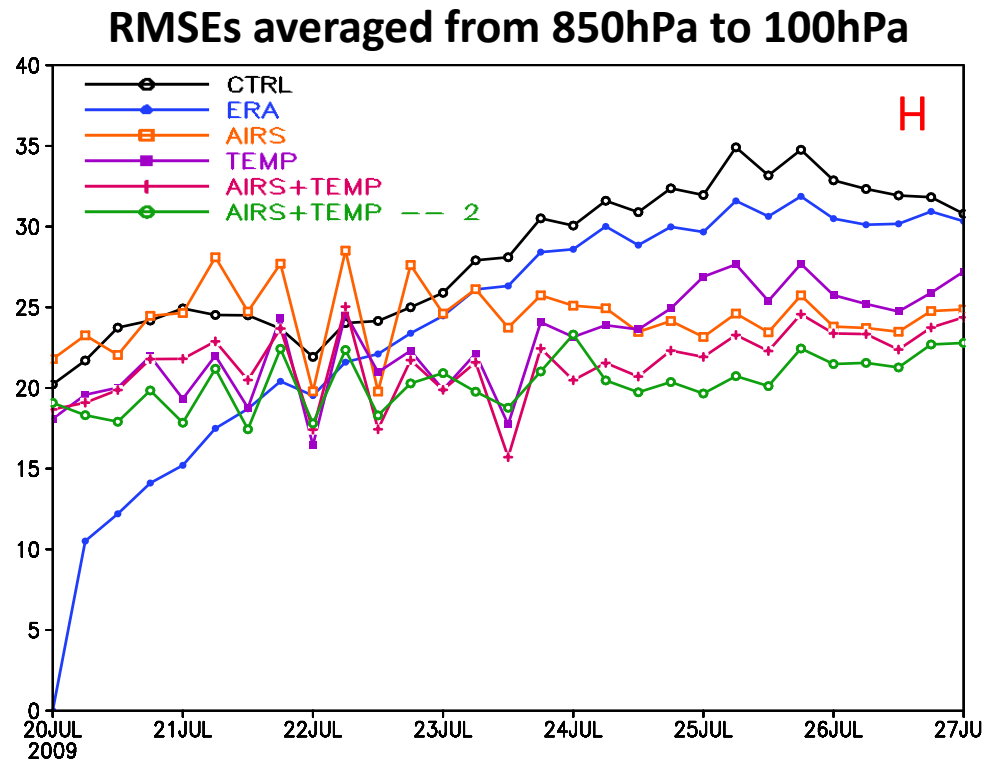
**ERA:** Using ERA40 as IC, no assimilation

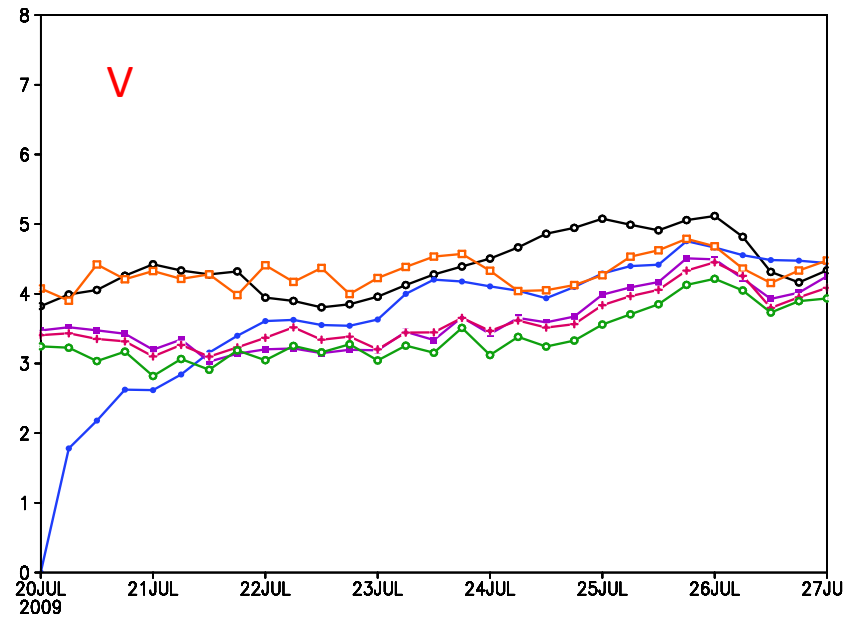
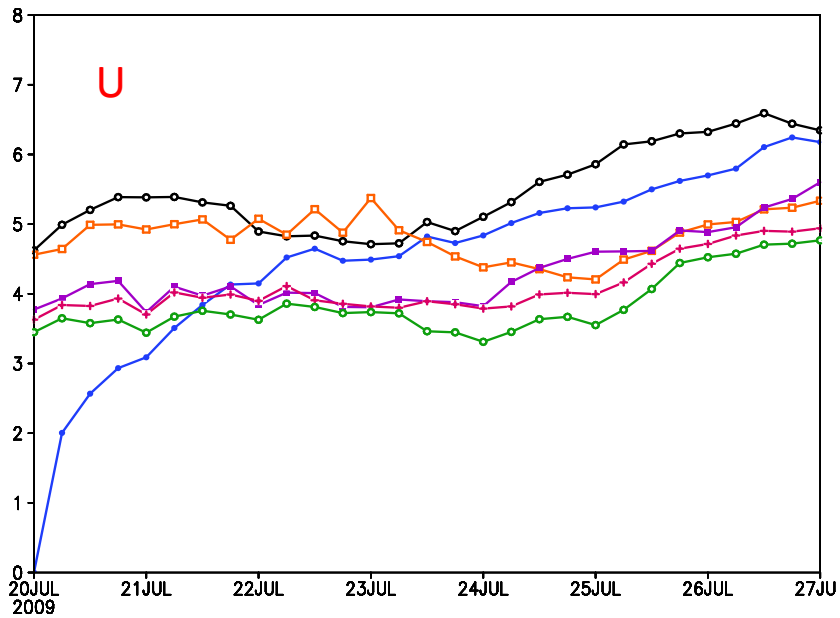
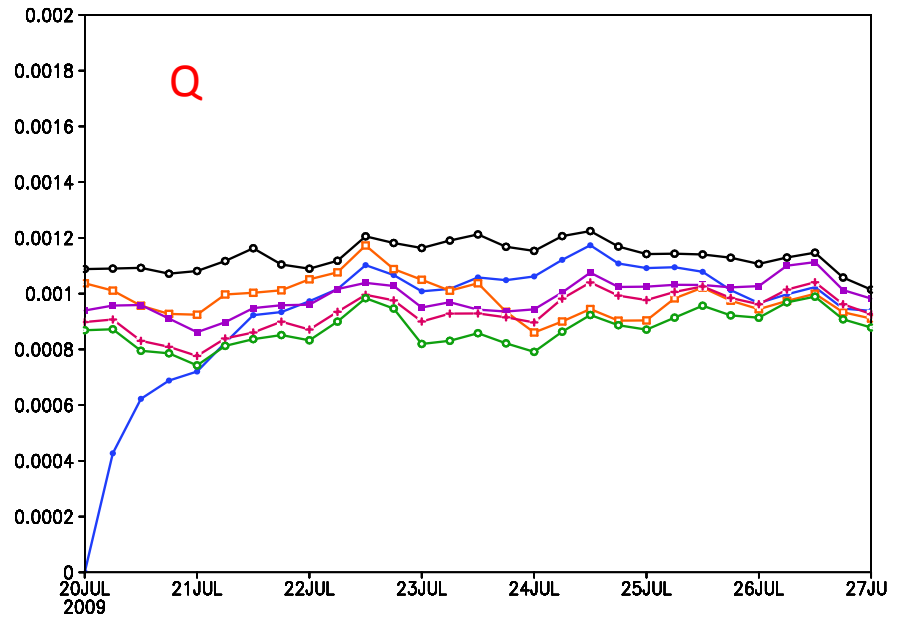
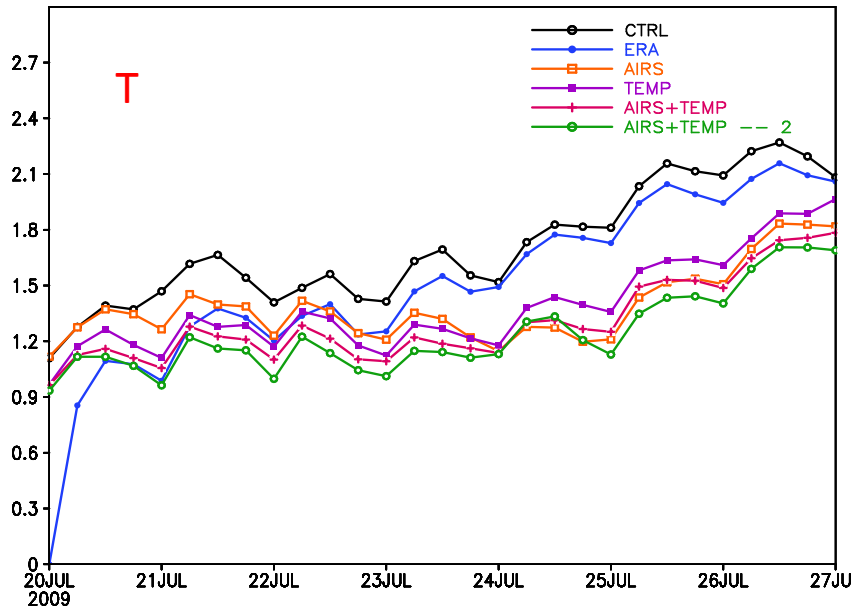
**AIRS:** Assimilating T & Q of AIRS only

**TEMP:** Assimilating Radiosondes only

**AIRS+TEMP:** With random perturbation

**AIRS+TEMP—2:** No random perturbation





# Summary

- ◆ **Localization for EnKF can be economically implemented using an orthogonal expansion of filter function;**
- ◆ **The new implementation can be applied to other ensemble-based DA, e.g., En-3DVar and En-4DVar.**



**Comments, please.**

**Thank you!**

- ◆ **“True” states:** simulations during a period of time unit after a long-term integration (e.g.,  $10^5$  model time steps) of the model from an arbitrary IC .
  - ◆ **Observations** of all model variables: “true” states plus uncorrelated random noise with standard Gaussian distribution (with zero mean and variance of 0.16)
  - ◆ **Assimilation experiments:** assimilation cycles over a period of 200 days using the **EnKF** with 18-hr assimilation window and, respectively.
- 500 samples** are used for the experiments so that they have good representativeness.