Development of an Ensemble Data Assimilation System for Mesoscale Ensemble Prediction

> Japan Meteorological Agency Numerical Prediction Division Tadashi Fujita

Meso ensemble prediction system

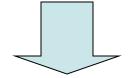
Objective: provide information on uncertainty in operational numerical prediction from the meso scale model (MSM)

JMA is testing various methodologies for possible operational use.

This presentation reports on our development of a system using LETKF(Local Ensemble Transform Kalman Filter (Hunt 2005))

Application of EnKF on JMANHM (non hydrostatic model used in operation): LETKF (Local Ensemble Transform Kalman Filter) : T. Miyoshi, and K. Aranami, 2006: SOLA, 2, 128-131.

positive impact in a perfect model experiment.



Development for possible operational use of the LETKF.

Current plan of JMA Meso ensemble prediction (may subject to change) resolution ~10km ensemble size ~ 5 members

Design of the LETKF analysis cycle

Analysis cycle at a lower resolution, but with a larger ensemble size

a large ensemble size is desirable for a stable analysis cycle ensemble forecast from selected members at a higher resolution

Analysis cycle on a larger domain than the forecast domain generate boundary and initial perturbations for ensemble forecast

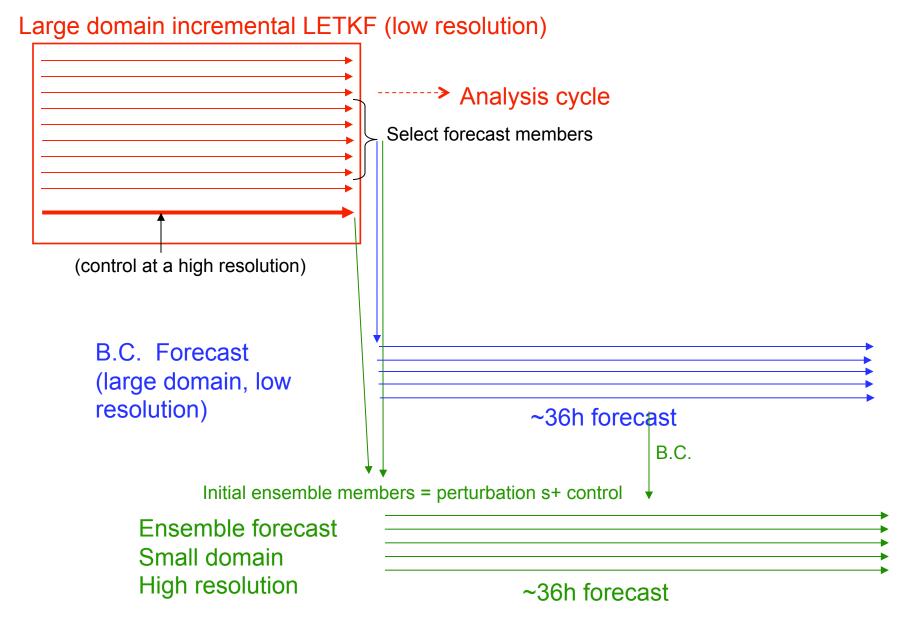
Consistent initial and boundary perturbations from a single LETKF system

Try to handle perturbations corresponding to severe events coming into the inner forecast domain through the boundary

(Boundary of the larger analysis domain currently is not perturbed.)

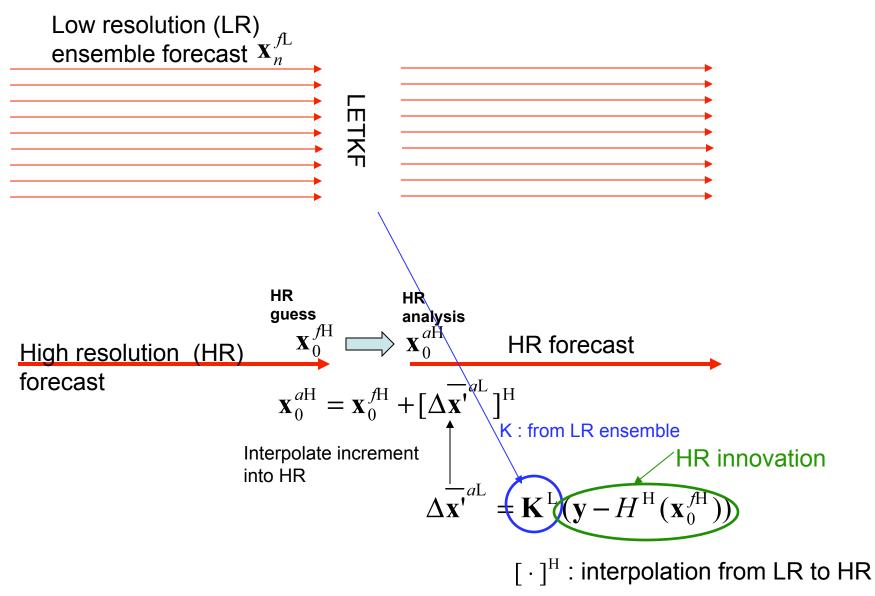
Incremental LETKF: low resolution ensemble + high resolution control

Dual Resolution EnKF (Gao and Xue, 2008: Mon. Wea. Rev., 136, 945-963.)



Incremental LETKF: Low resolution ensemble + High resolution control

Dual Resolution EnKF (Gao and Xue, 2008: Mon. Wea. Rev., 136, 945-963.)



LETKF analysis cycle experiment

24 Aug. 2008.08.24 12UTC - 28 Aug. 2008 00UTC

LR:40km 50 members HR:20km

Larger domain: 20km , 281x305 grid points

50 vertical levels

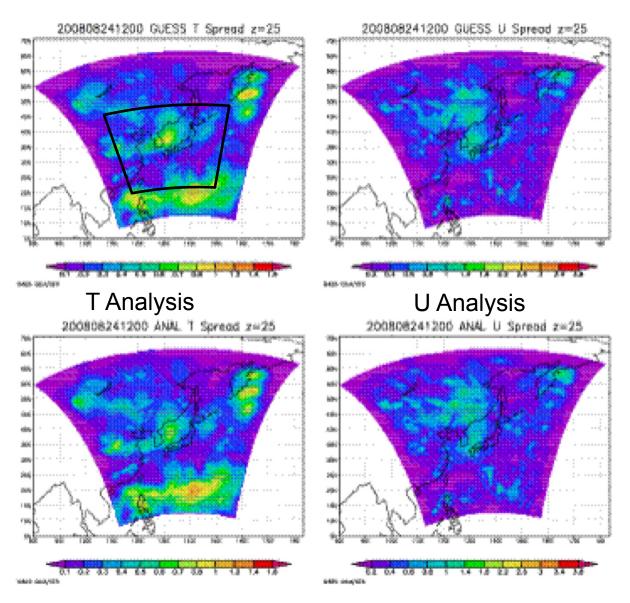
Observational data: conventional data + Doppler velocity Observation operator from operational meso analysis (4DVar)

Localization scale: horrizontal 5 grid points (200km), vertical 3 levels

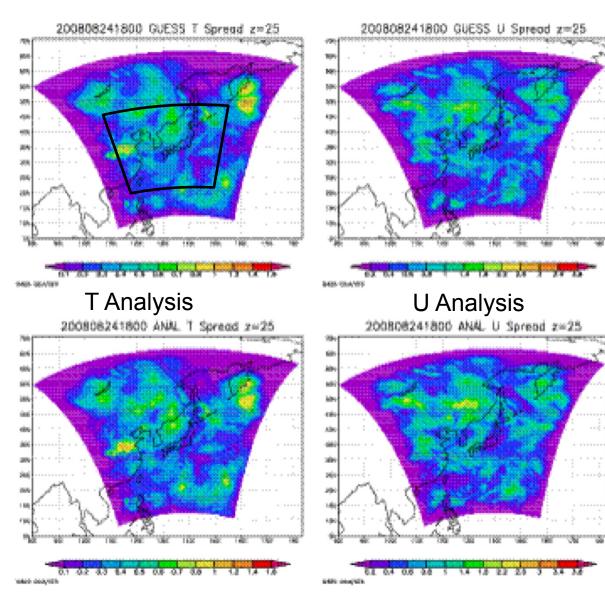
Assimilation window 3h

Forecast 28 Aug. 00UTC – 29 Aug. 12UTC 20km 181x145x50 grid points

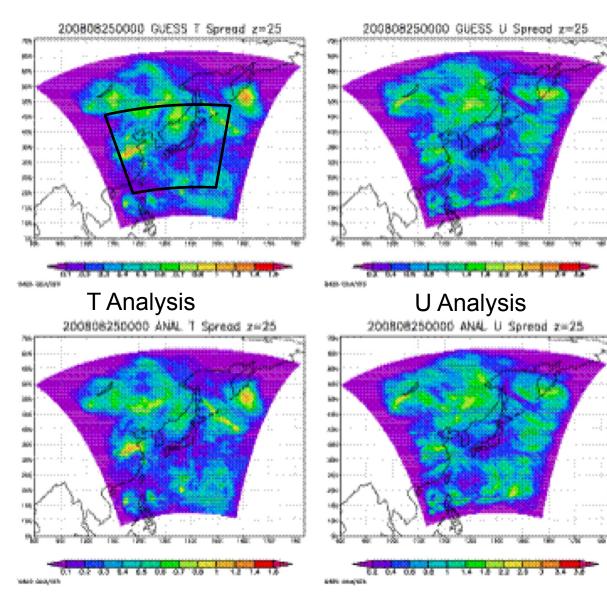
T Guess



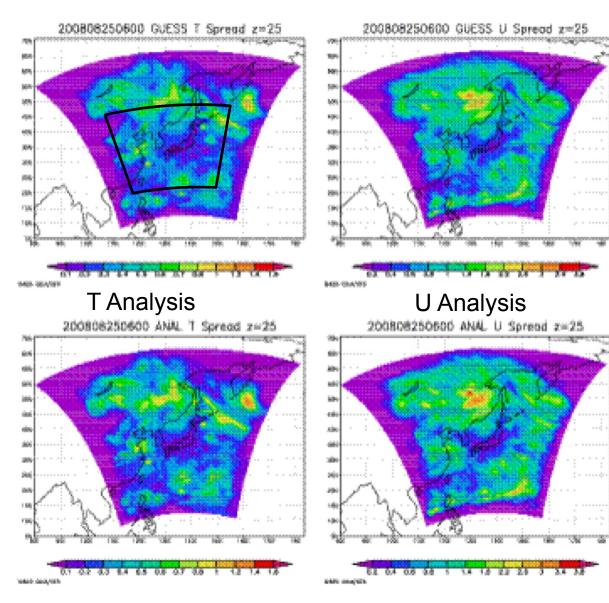
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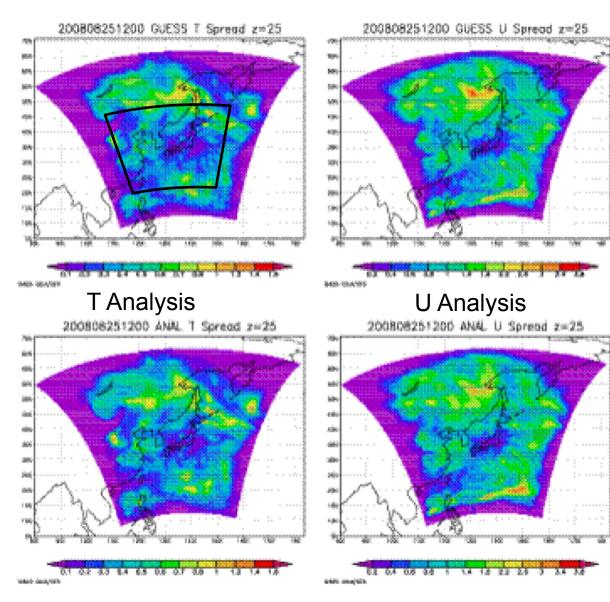
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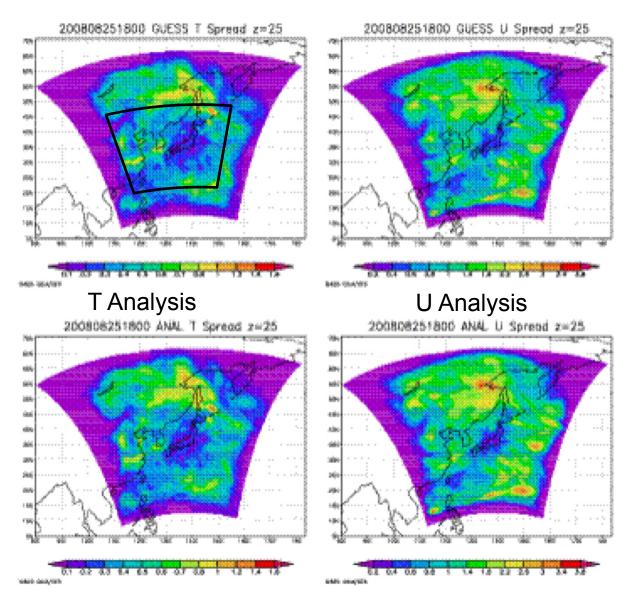
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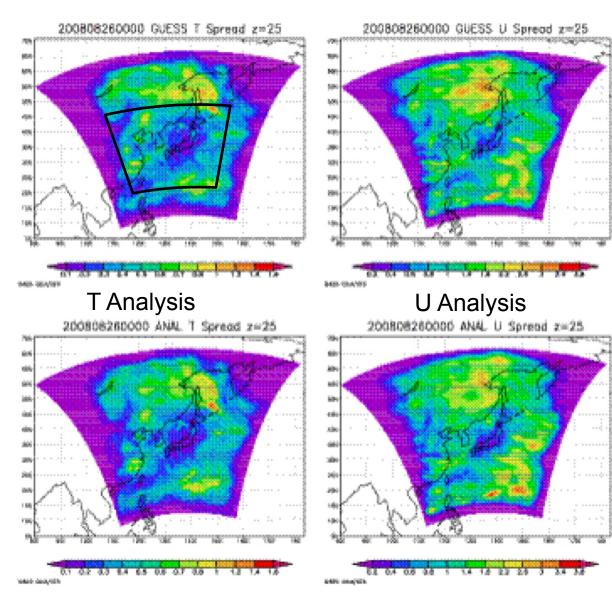
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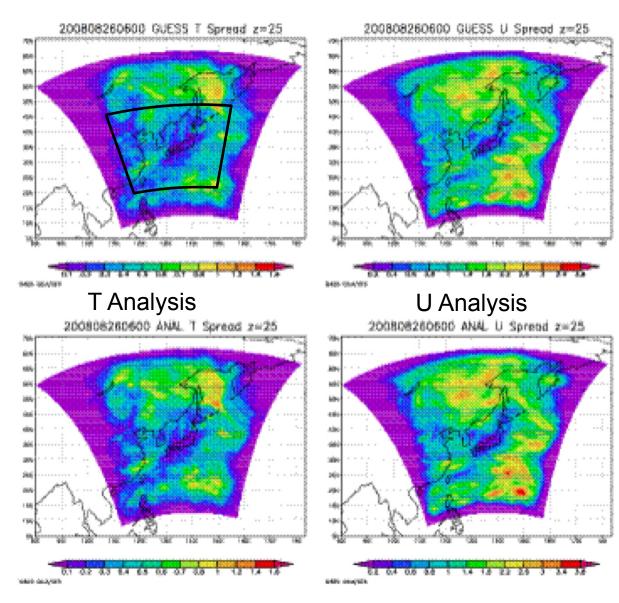
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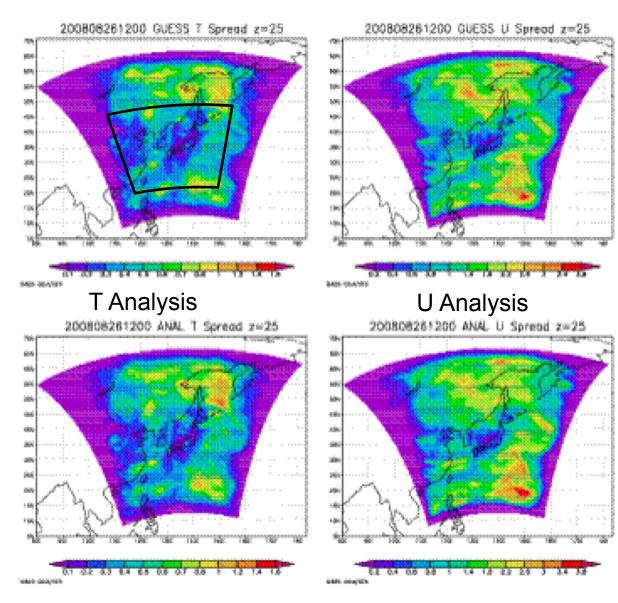
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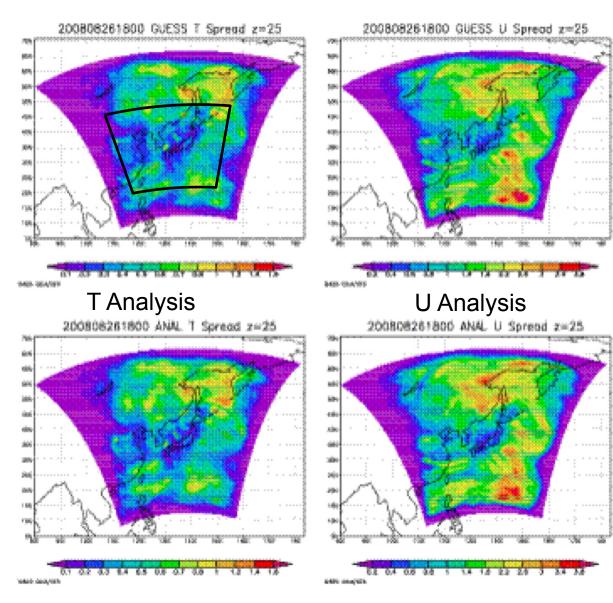
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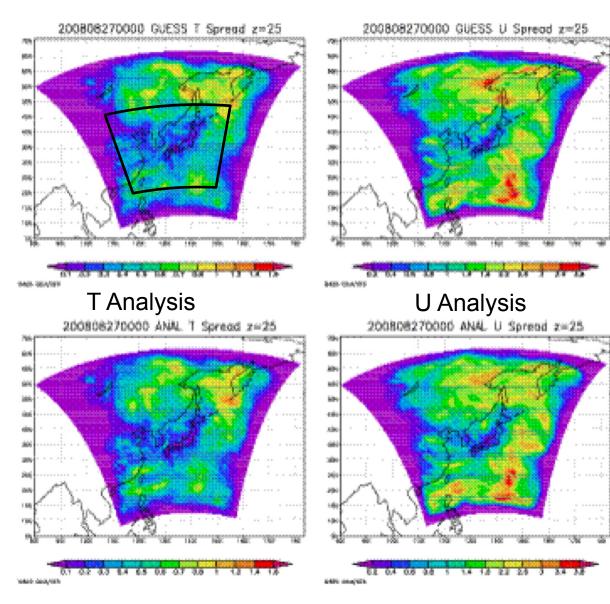
T Guess



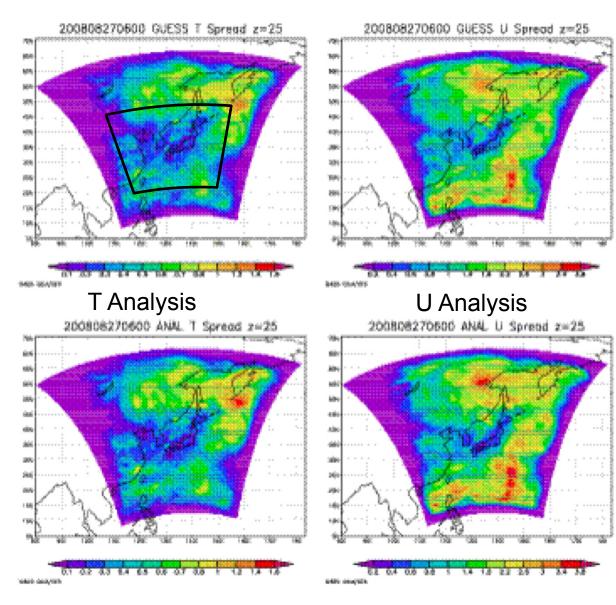
T Guess



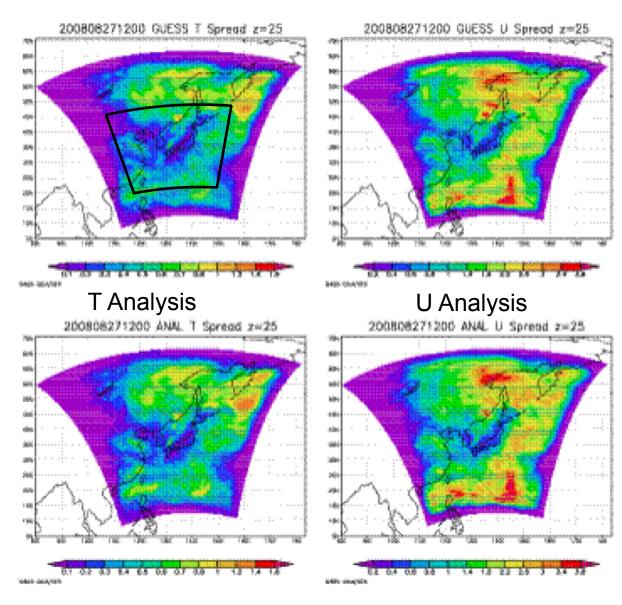
T Guess



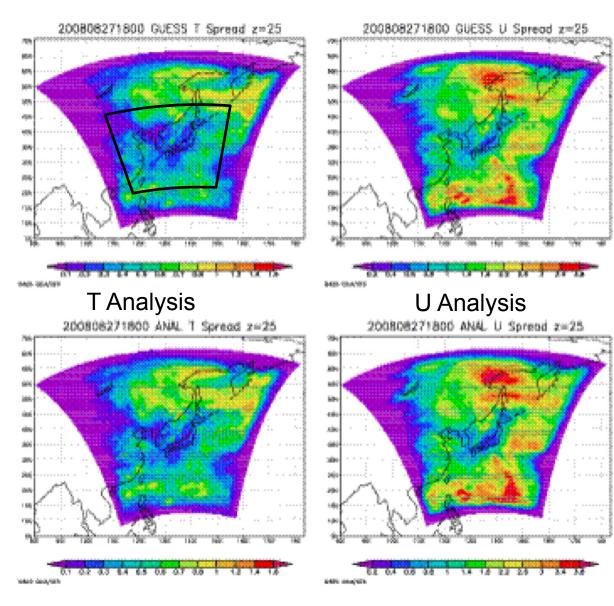
T Guess



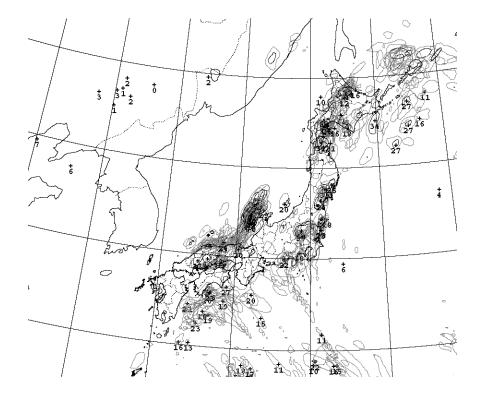
T Guess



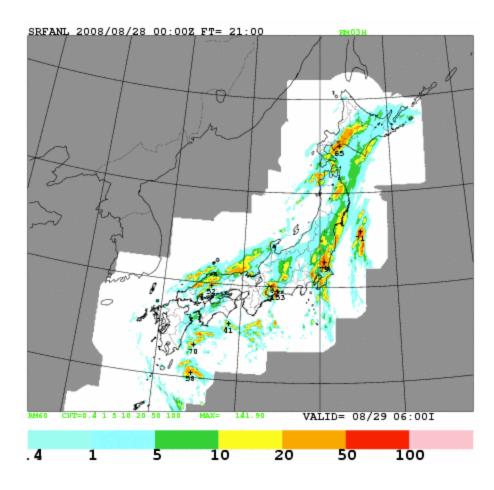
T Guess



Ensemble forecast from the LETKF analysis FT=21 (30 members) gray:10mm/3h, black: 20mm/3h



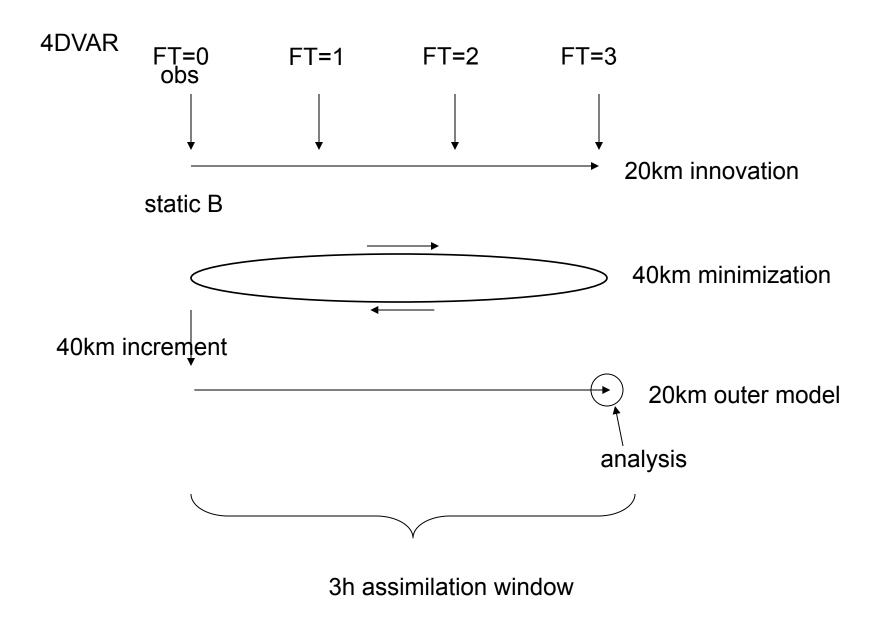
28 Aug. 21UTC3h Accumulated precipitation

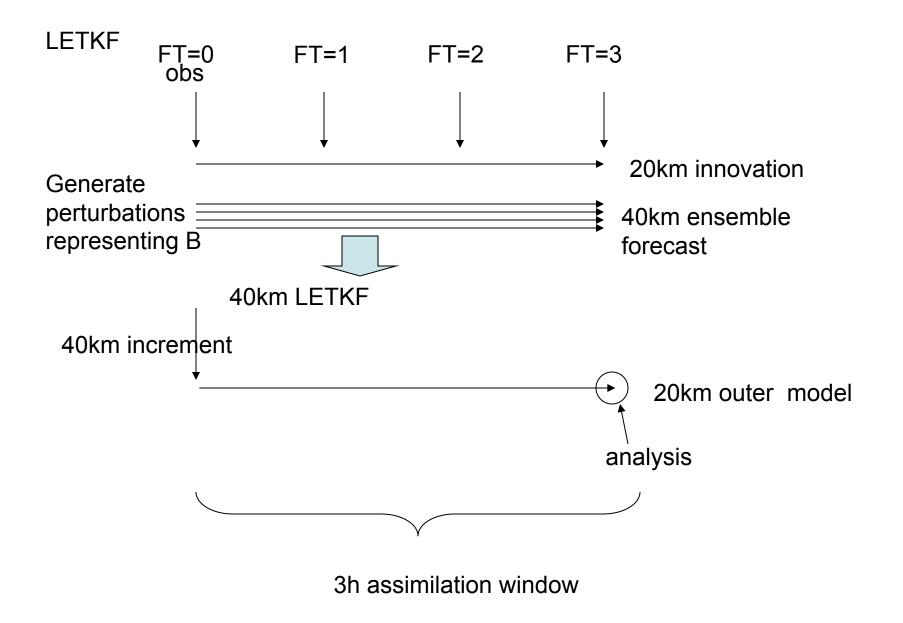


LETKF with static B

LETKF using perturbations based on Background error covariance in variational method.

Compare LETKF and 4DVAR under the same condition (assimilation window, resolution, innovation)



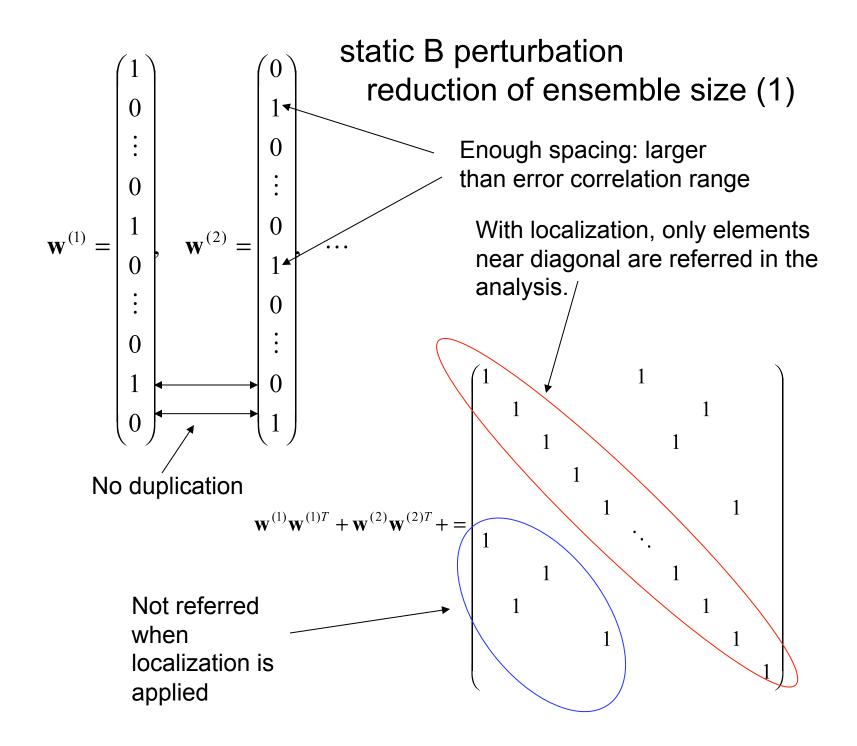


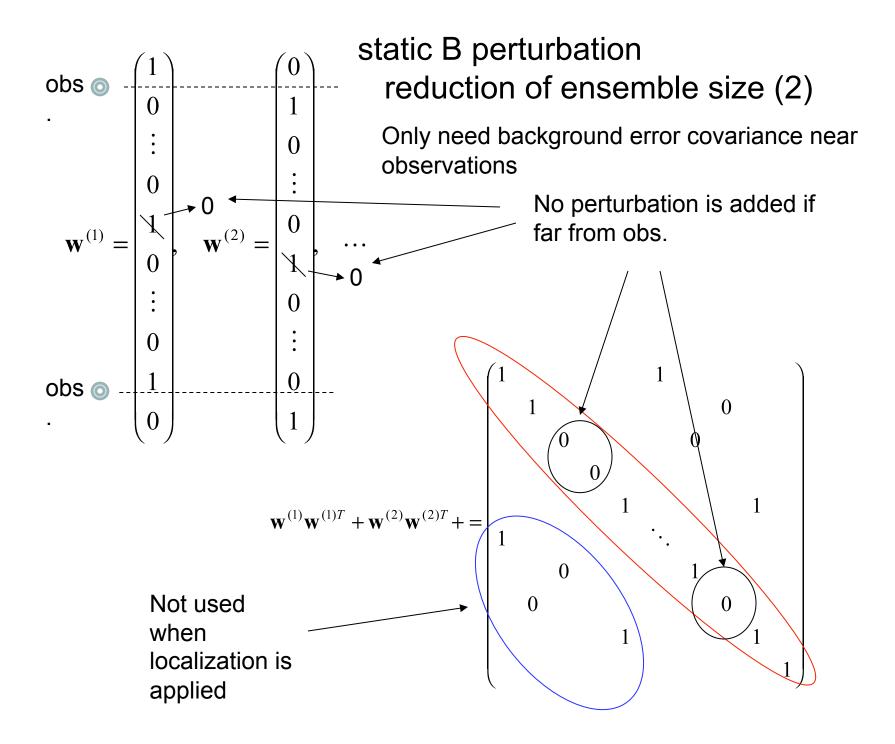
Ensemble covariance

$$\mathbf{P} = \mathbf{E}\mathbf{E}^{T}$$
$$\mathbf{E} = \frac{1}{\sqrt{m-1}} [\delta \mathbf{x}^{(1)} | \cdots | \delta \mathbf{x}^{(m)}]$$

Static covariance

 $\mathbf{B} = \mathbf{F}\mathbf{F}^T \qquad \mathbf{F} = [\mathbf{v}^{(1)} | \cdots | \mathbf{v}^{(m)}]$ $\mathbf{v}^{(l)} = \mathbf{B}^{1/2} \mathbf{w}^{(l)}$ $\mathbf{w}^{(1)} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad \mathbf{w}^{(2)} = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \end{pmatrix}, \quad \cdots \quad \mathbf{w}^{(1)} \mathbf{w}^{(1)T} + \mathbf{w}^{(2)} \mathbf{w}^{(2)T} + \cdots = 1$ Need huge ensemble size (equal to number of degrees of freedom of the system) => reduce ensemble size





$$\mathbf{v}^{(l)} = \mathbf{B}^{1/2} \mathbf{w}^{(l)} \quad \text{correct formulation, but not balanced}$$
$$v_i^{(l)} = \sum_j \left\langle (\Delta x_i) (\Delta x_j / \boldsymbol{\sigma}_{x_j})^T \right\rangle w_j^{(l)}$$

Procedure to generate perturbation

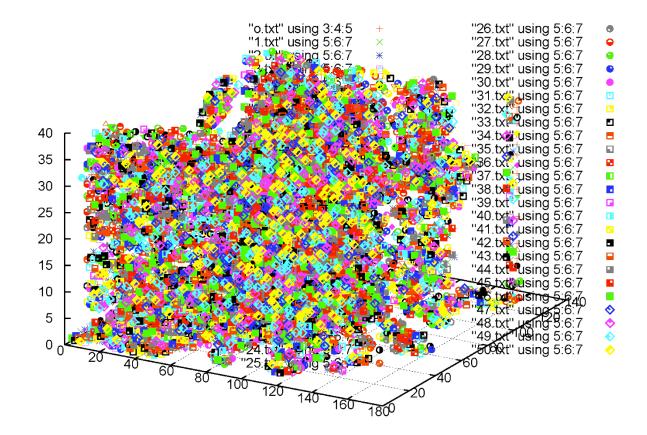
- 1. Check observation distribution
- 2. Select grid points near obs.(apply thinning)
- 3. Distribute selected grid points to ensemble members
- 4. Put +1, -1, +1, -1 (seeds) on selected grid points.
- 5. Apply B

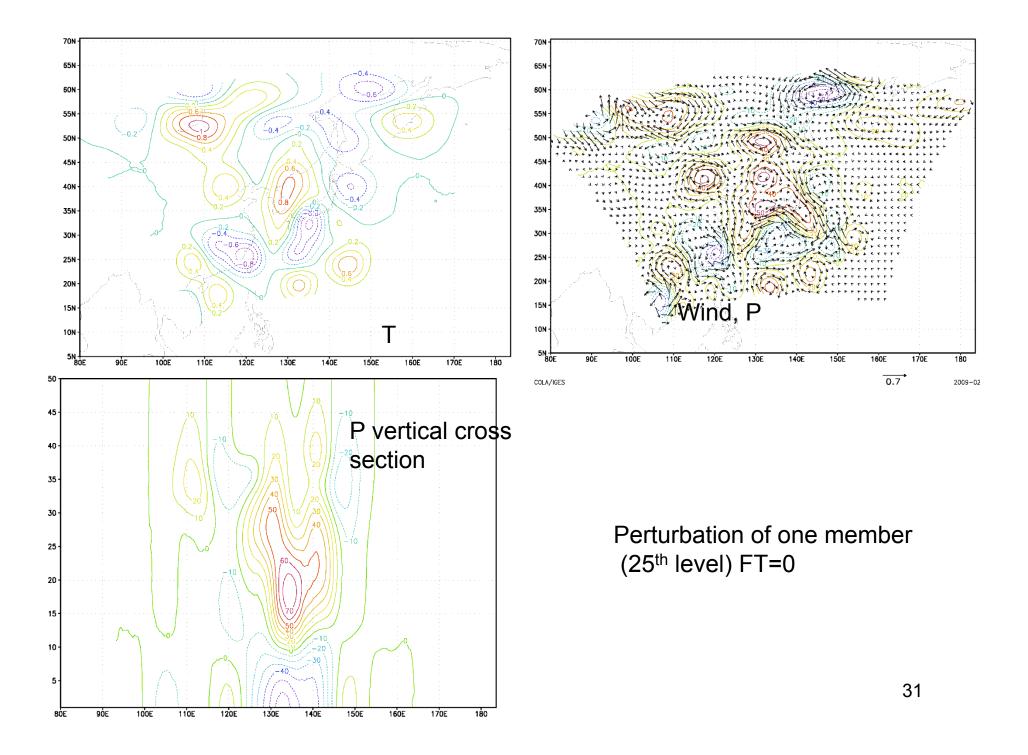
LETKF pert distribution of seeds of the perturbations

Red: observation points "o.txt" using 3:4:5 "1.txt" using 5:6:7 +Green : perturbation seeds \times of one member $\times +$ 0 0**'** 1つ

Distribution of perturbation seeds

(color indicates member)

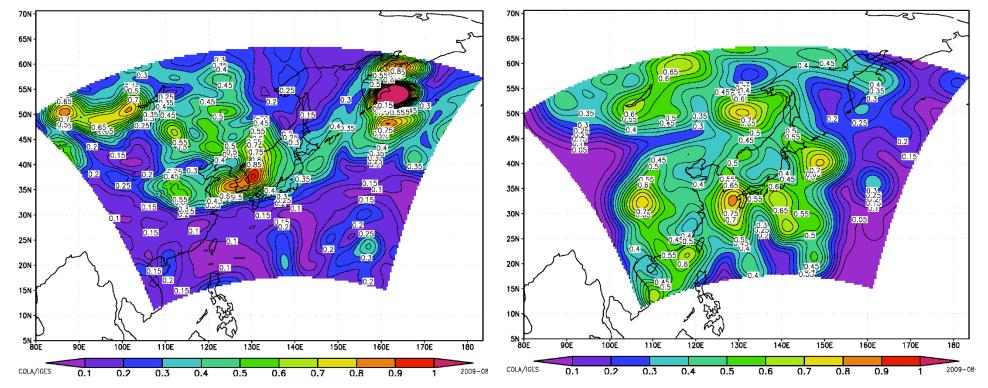




T spread 25th level

Operational global ensemble forecast 50 members

Generated perturbation



4DVAR (hydro static, (outer model non-hydrostatic)) Control variables u_u, v_u, T_v, P_S, q

Incremental method (inner 40km, outer 20km)

LETKF using the generated Var. perturbations

 u_u, v_u, T_v perturbation (seeded around observation points)

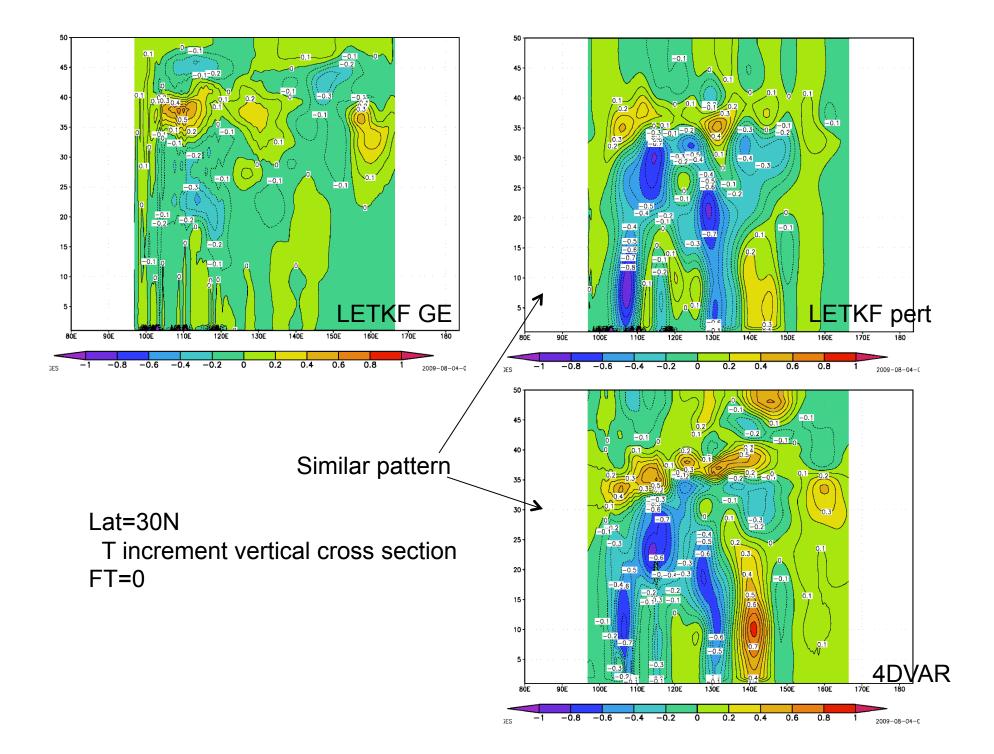
±7 grid points, every 2 grid points ±5 vertical levels, every 3 levels seeds distributed different members (25 members for each element) => 75 members

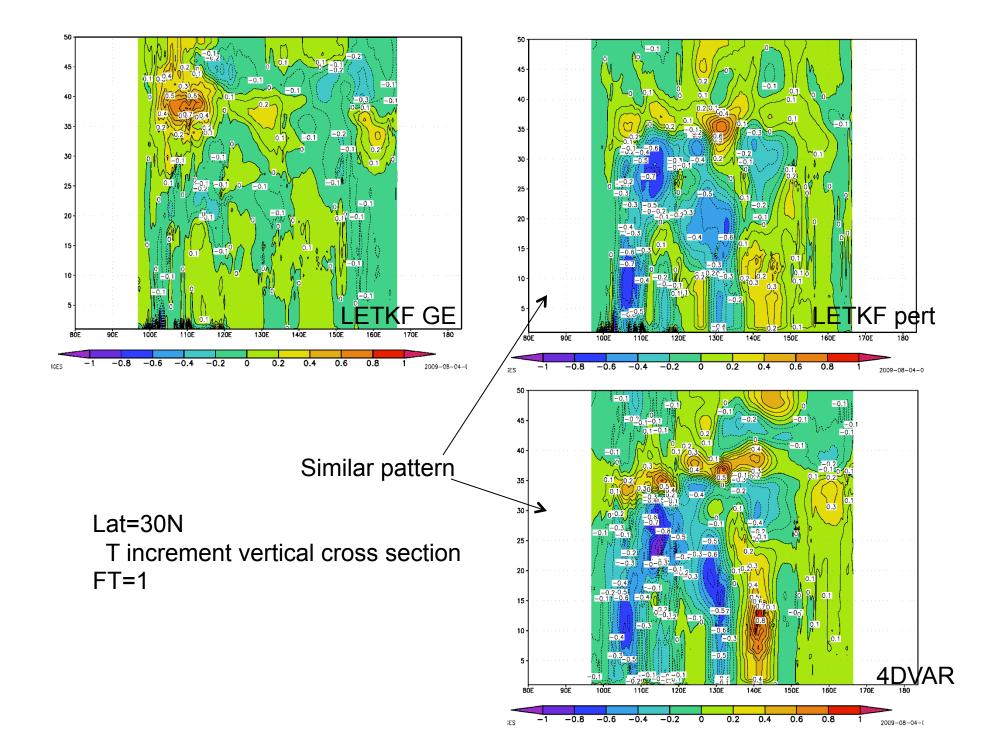
Incremental method (40km ensemble + 20km control)

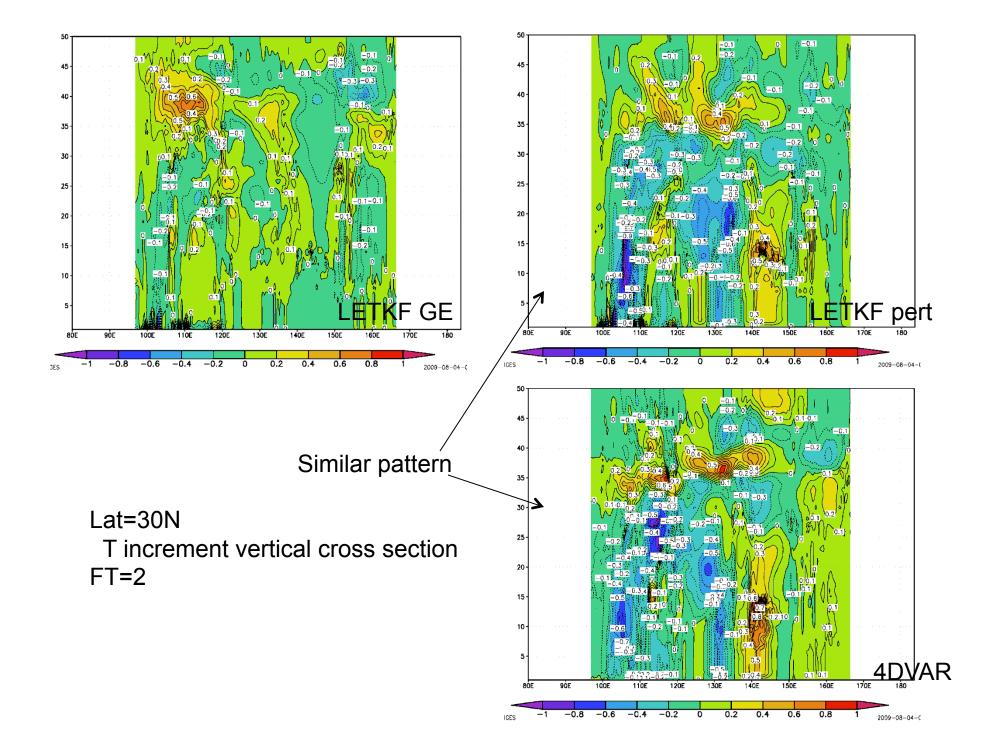
Localization scale: horrizontal 10 grid points, 8 vertical levels

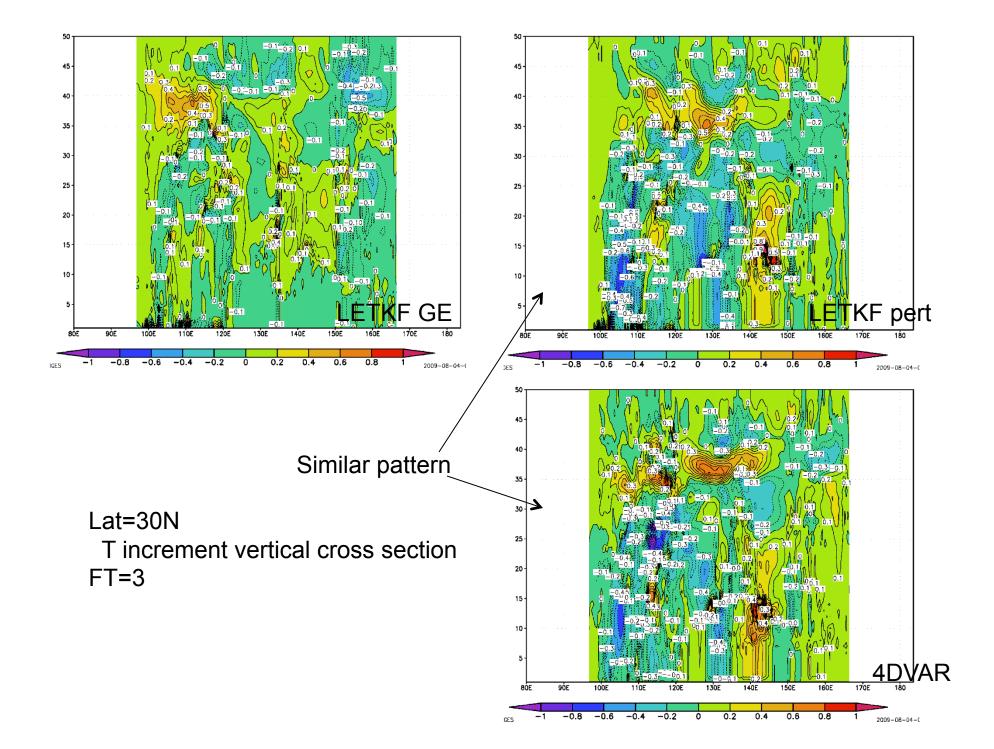
LETKF GE

LETKF using perturbations from operational global ensemble forecast (SV, 50 members)

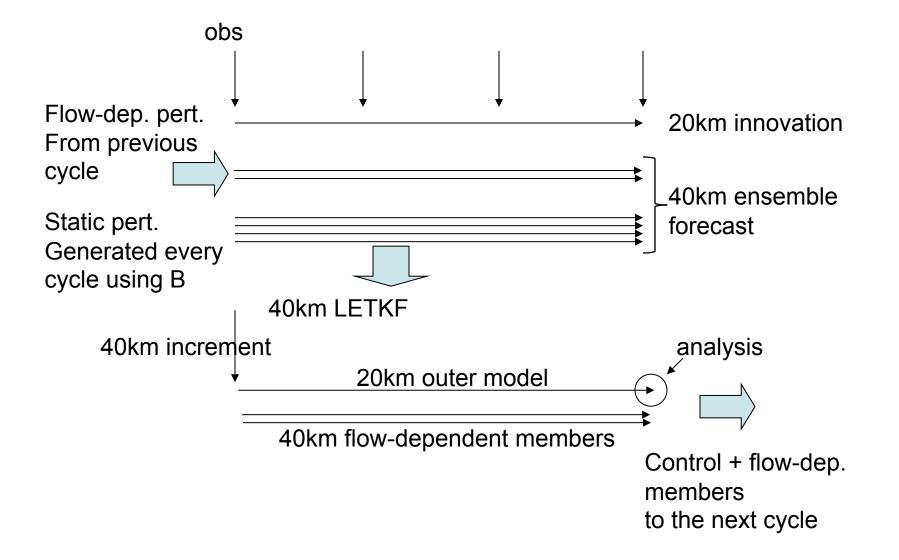








Possible design of a system using flow-dependent and static perturbations



Using perturbations with different scales in ensemble data assimilation (test with variational method, displacement perturbations)

test to handle errors in different scales in data assimilation.

Variational framework Using ensemble perturbations in variational method

$$\mathbf{J} = \frac{1}{2} \mathbf{v}_{0}^{T} \mathbf{v}_{0} + \frac{1}{2} \mathbf{v}_{pert}^{T} \mathbf{v}_{pert} \qquad \text{Hamill and Snyder, 2000} \\ \text{Lorenc, 2003} \\ \text{Buehner et al., 2004} \\ \text{Wang et al., 2008} \\ + \frac{1}{2} (\mathbf{y} - H(\mathbf{x}^{b} + \Delta \mathbf{x}))^{T} \mathbf{R}^{-1} (\mathbf{y} - H(\mathbf{x}^{b} + \Delta \mathbf{x})) \\ \Delta \mathbf{x} = \beta^{1/2} \mathbf{B}_{0}^{1/2} \mathbf{v}_{0} + \sqrt{1 - \beta} \mathbf{B}_{pert}^{1/2} \mathbf{v}_{pert} \qquad \text{Localization (allow to} \\ \text{specify localization of} \\ \text{static} \qquad \text{ensemble} \qquad \qquad \text{Localization (allow to} \\ \mathbf{y} = \mathbf{p}_{pert}^{1/2} \mathbf{B}_{0}^{1/2} \mathbf{v}_{0} + \sqrt{1 - \beta} \mathbf{B}_{pert}^{1/2} \mathbf{v}_{pert} \\ \mathbf{B}_{pert}^{1/2} = \frac{1}{\sqrt{M - 1}} (\mathbf{D}_{1} \mathbf{L}_{1}^{1/2} \mathbf{D}_{2} \mathbf{L}_{2}^{1/2} \cdots \mathbf{D}_{M} \mathbf{L}_{M}^{1/2}) \\ \mathbf{v}_{pert} : \text{NxM dimensional vector} \\ \text{(weight of each member on each grid point)} \\ \mathbf{D}_{m} = \begin{pmatrix} \mathbf{D}_{m}^{1} \\ \mathbf{D}_{m}^{2} \\ \vdots \\ \mathbf{D}_{m}^{L} \end{pmatrix} \qquad \mathbf{D}_{m}^{I} = \begin{pmatrix} \mathbf{\Delta}_{m(l,1)} \\ \mathbf{\Delta}_{m(l,2)} \\ \vdots \\ \mathbf{\Delta}_{m(l,N_{space})} \end{pmatrix}$$

Simple example of perturbation horizontal displacement perturbation (scale selective)

Apply smoothing: Pick up information of specified scale

$$\overline{\mathbf{x}}_{\Delta}(i,j,k) = \frac{1}{\left(2\Delta+1\right)^2} \sum_{i'=-\Delta}^{\Delta} \sum_{j'=-\Delta}^{\Delta} \mathbf{x}(i+i',j+j',k)$$

Generate perturbations from x- and y- displacement Consider only uncertainties from horizontal displacement error

$$\delta \mathbf{x}_{x}^{\Delta}(i, j, k) = \overline{\mathbf{x}}_{\Delta}(i + \Delta/2, j, k) - \overline{\mathbf{x}}_{\Delta}(i - \Delta/2, j, k)$$

$$\delta \mathbf{x}_{y}^{\Delta}(i, j, k) = \overline{\mathbf{x}}_{\Delta}(i, j + \Delta/2, k) - \overline{\mathbf{x}}_{\Delta}(i, j - \Delta/2, k)$$

*Smooth with scale delta to pick up displacement error in the specified scale. *use several deltas (error correlation in different scales)

* x- and y- perturbations (independent 2 directions) represent horizontal displacement in every direction.

* Assign larger localization scale for larger scale error (correlation signal reaches over longer distance)

cf. Aonashi and Eito, J. Meteor. Soc. Japan 2010

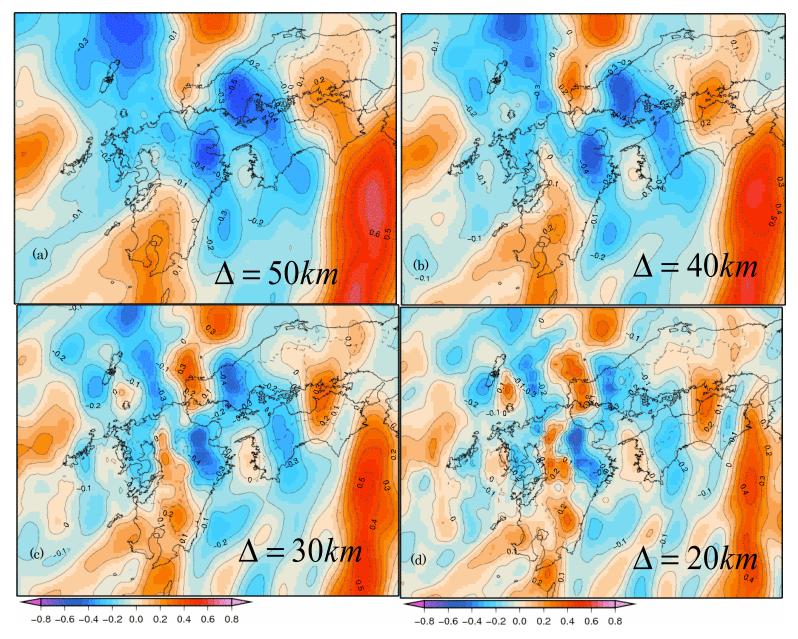
Example (5km grid spacing)

using 4 scales => x- and y- directions x 4 scales = 8 members

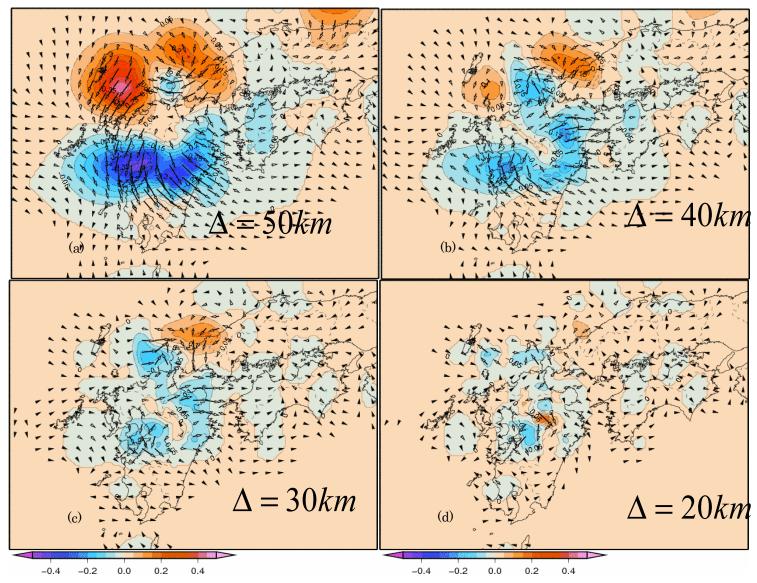
Assign independent degrees of freedom of control variable to each scale and each direction at each grid point => v_pert : 8 x N_space dimensional vector



Potential temperature x-displacement perturbation 25th level (~ 500hPa)



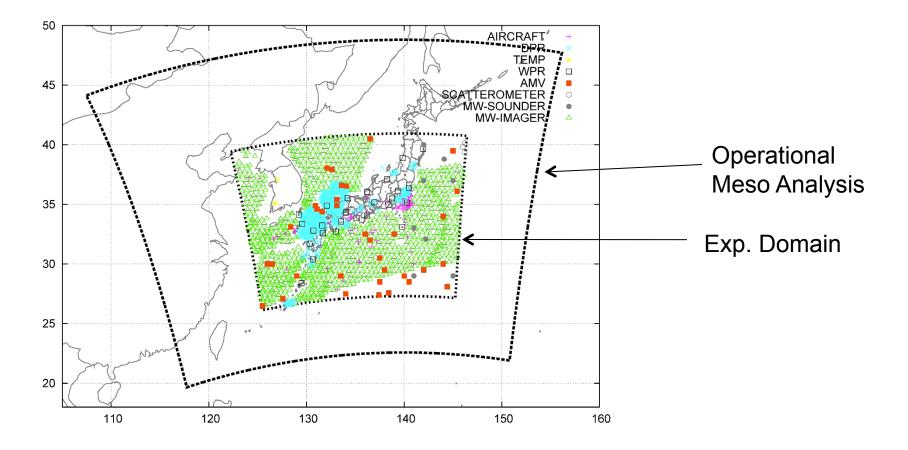
Potential temperature increment, displacement vector 25th level (~ 500hPa)

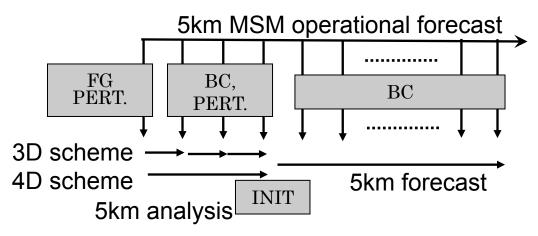


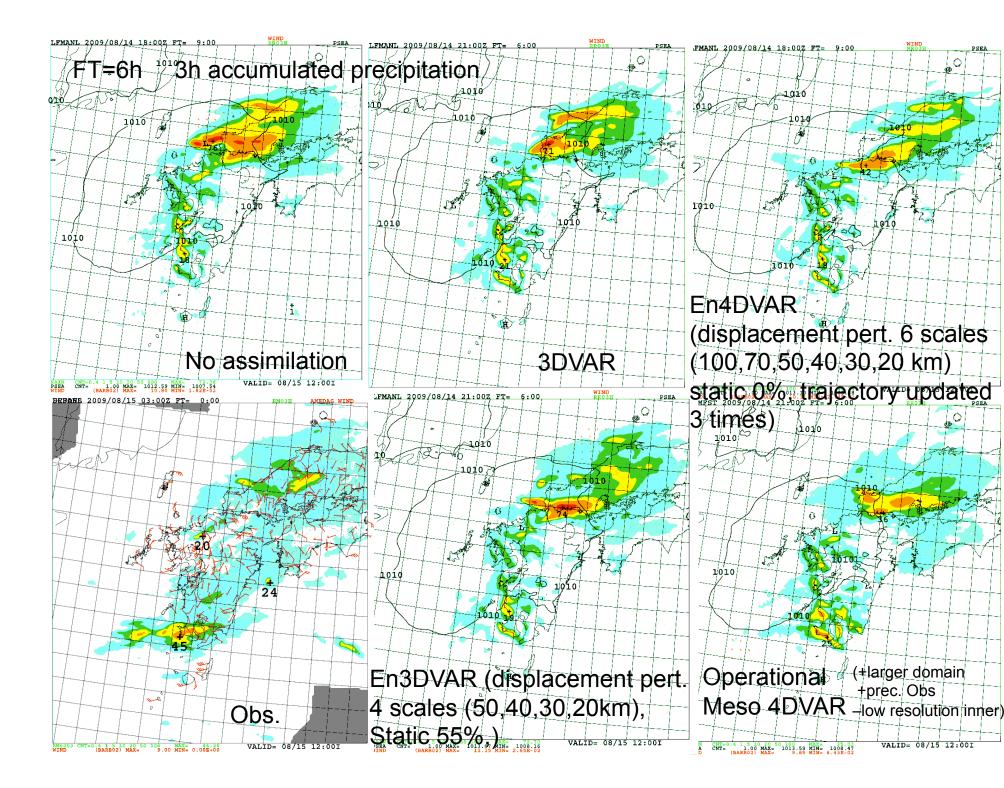
4 dimensional generalization of the formulation

$$J = \frac{1}{2} \mathbf{v}_{0}^{T} \mathbf{v}_{0} + \frac{1}{2} \mathbf{v}_{pert}^{T} \mathbf{v}_{pert} \qquad \begin{array}{c} \text{Liu et al. MWR 2008, 2009} \\ \text{Buehner et al. MWR 2009} \\ + \frac{1}{2} (\mathbf{y} - H(\mathbf{x}^{b} + \Delta \mathbf{x}))^{T} \mathbf{R}^{-1} (\mathbf{y} - H(\mathbf{x}^{b} + \Delta \mathbf{x})) \\ HM \\ \mathbf{x}^{b} \Longrightarrow M \mathbf{x}^{b} \\ \delta \mathbf{x}_{x}^{\Delta}(i, j, k) \Longrightarrow \overline{M} \mathbf{x}_{\Delta}(i + \Delta/2, j, k) - \overline{M} \mathbf{x}_{\Delta}(i - \Delta/2, j, k) \end{array}$$

Use time correlation of perturbations Instead of tangent linear and adjoint models







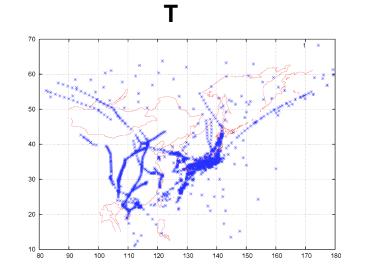
Summary

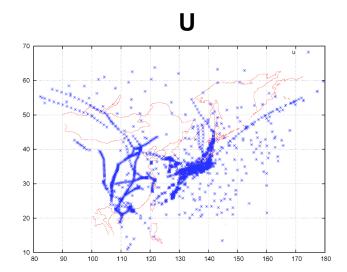
Meso scale LETKF analysis cycle system is developed. A cycle analysis experiment is performed.

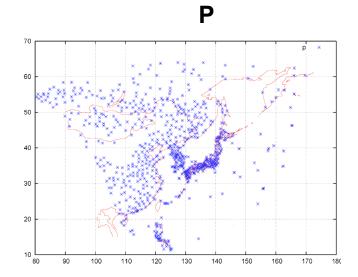
Perturbations based on static B from Var. are tested in LETKF. May work to make analysis similar to 4DVAR. (may help spin-up of analysis cycle, may mitigate influence of model error)

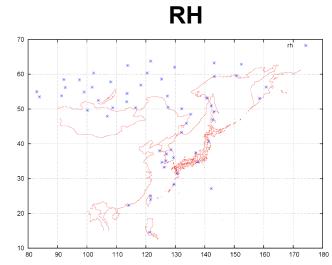
Property of ensemble perturbations is important in LETKF. possibility of designing perturbations with different features (static/ flow-dep., different scales)

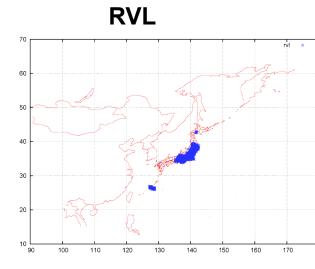
25 Aug. 2008 00-03UTC observations

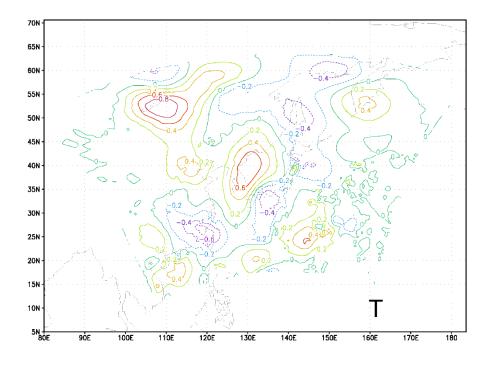


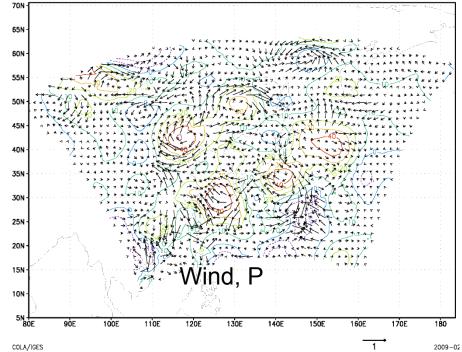


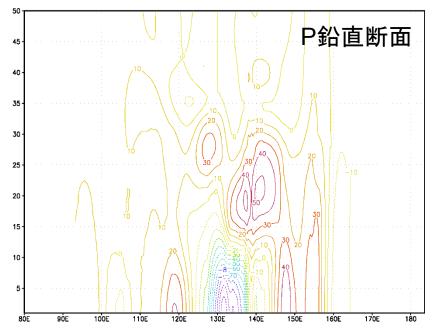






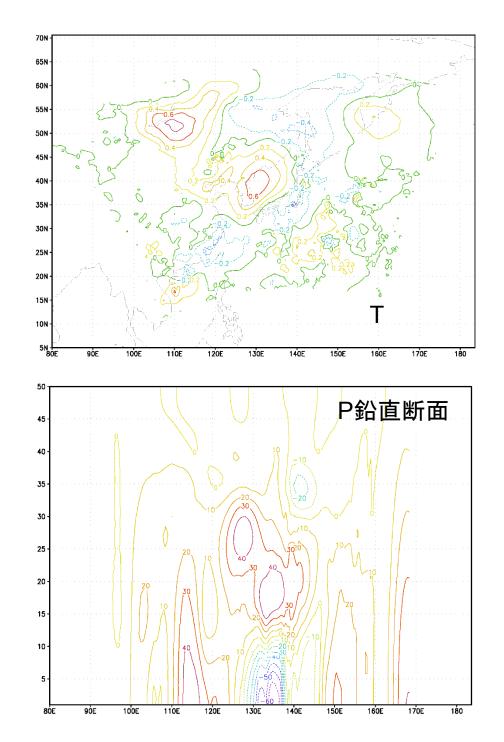


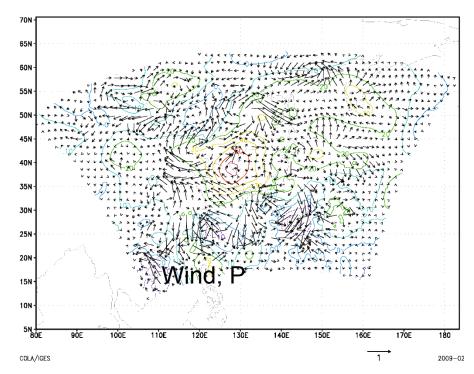




あるメンバーの摂動の時 間発展(25層)

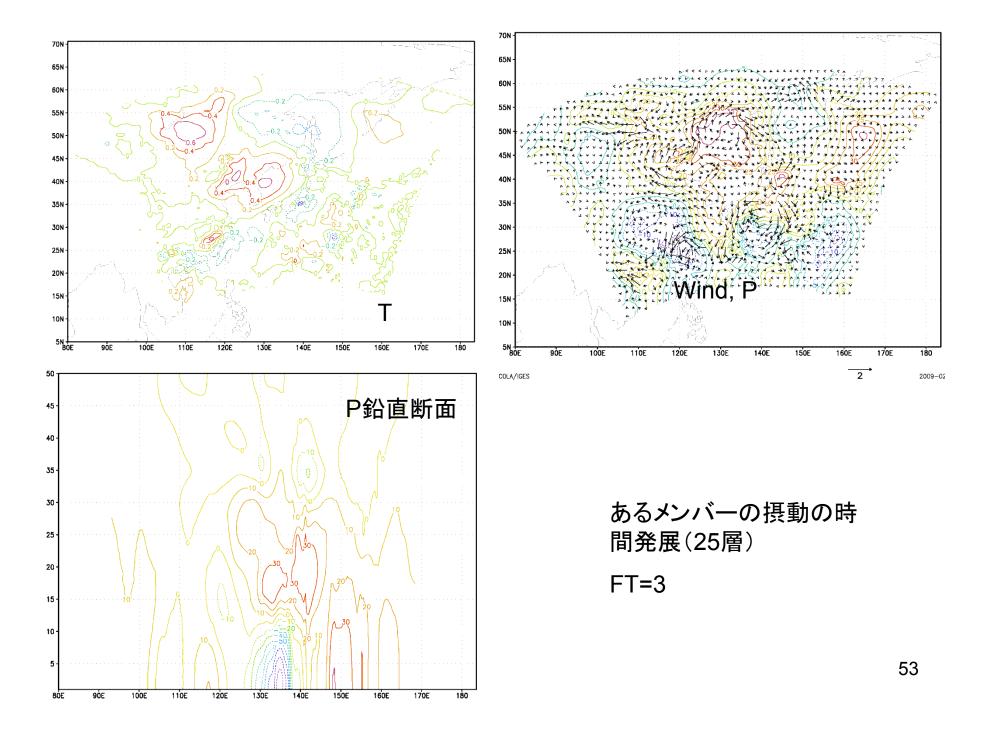
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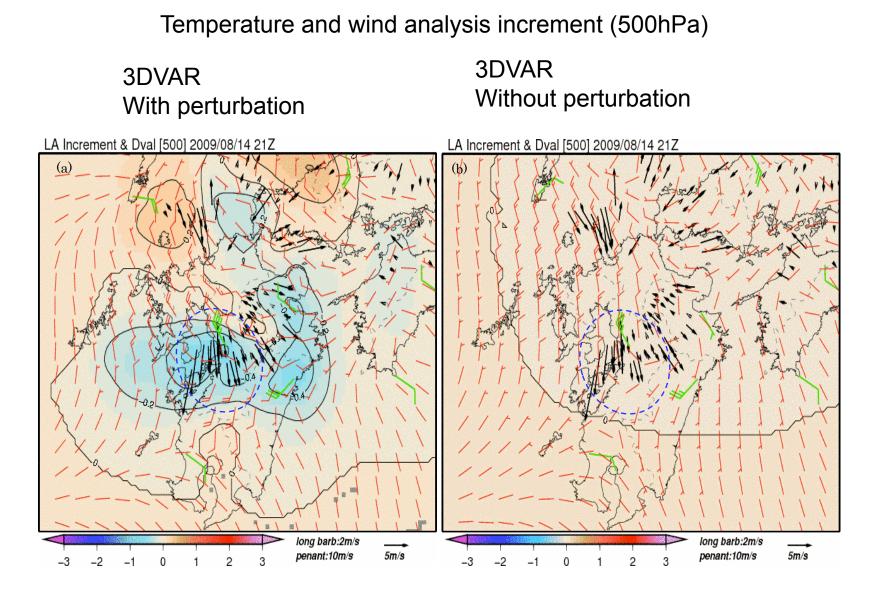


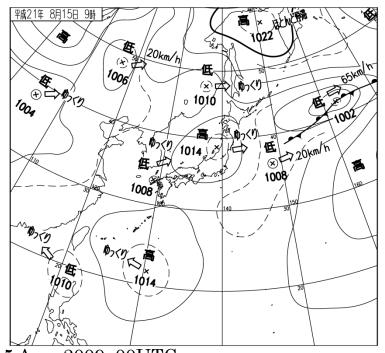


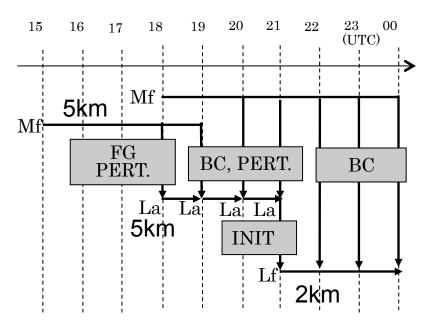
あるメンバーの摂動の時 間発展(25層)

FT=2

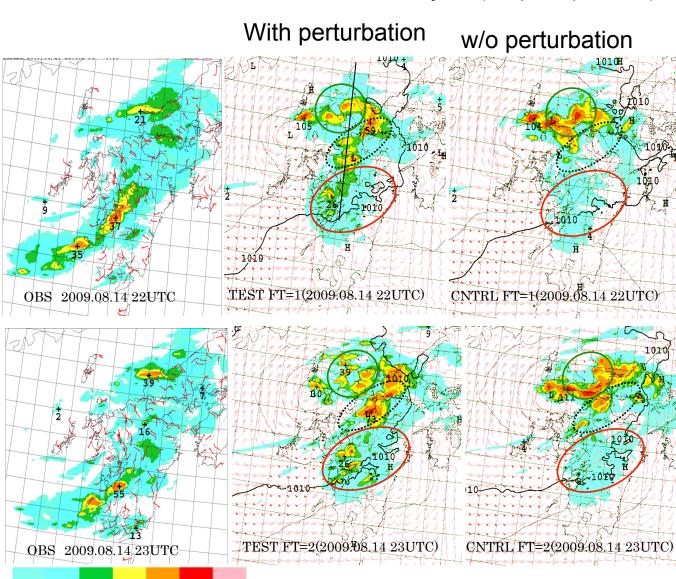








15 Aug. 2009, 00UTC



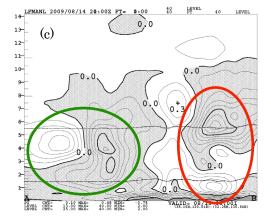
0.4

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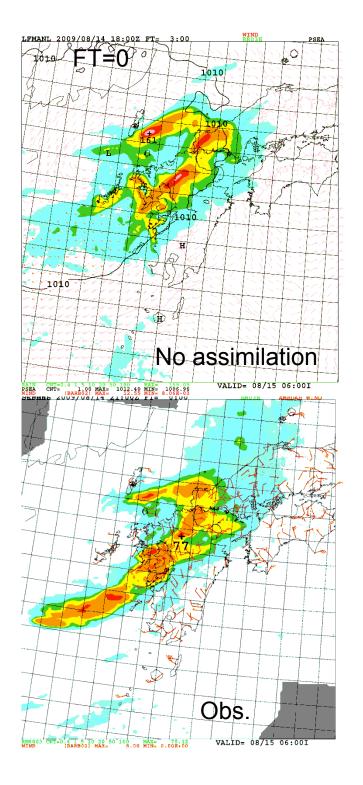
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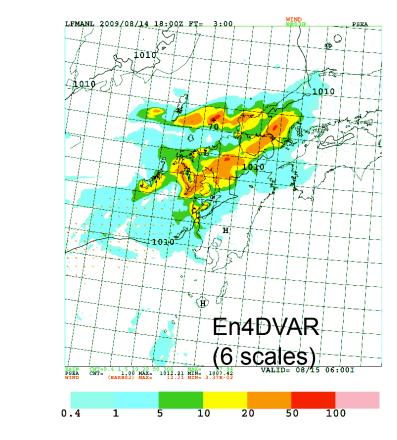
50

2km forecast from the analysis (1h precipitation)



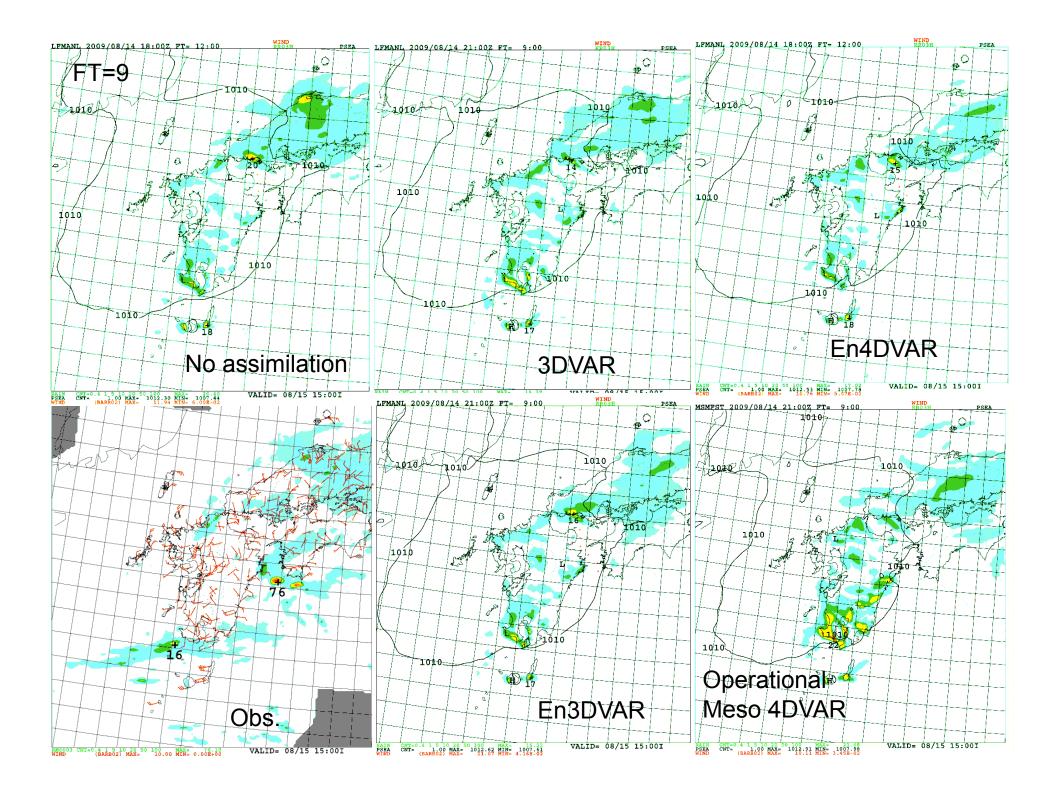
Potential temperature increment vertical cross section (contoured every 0.1K, Shaded: negative)

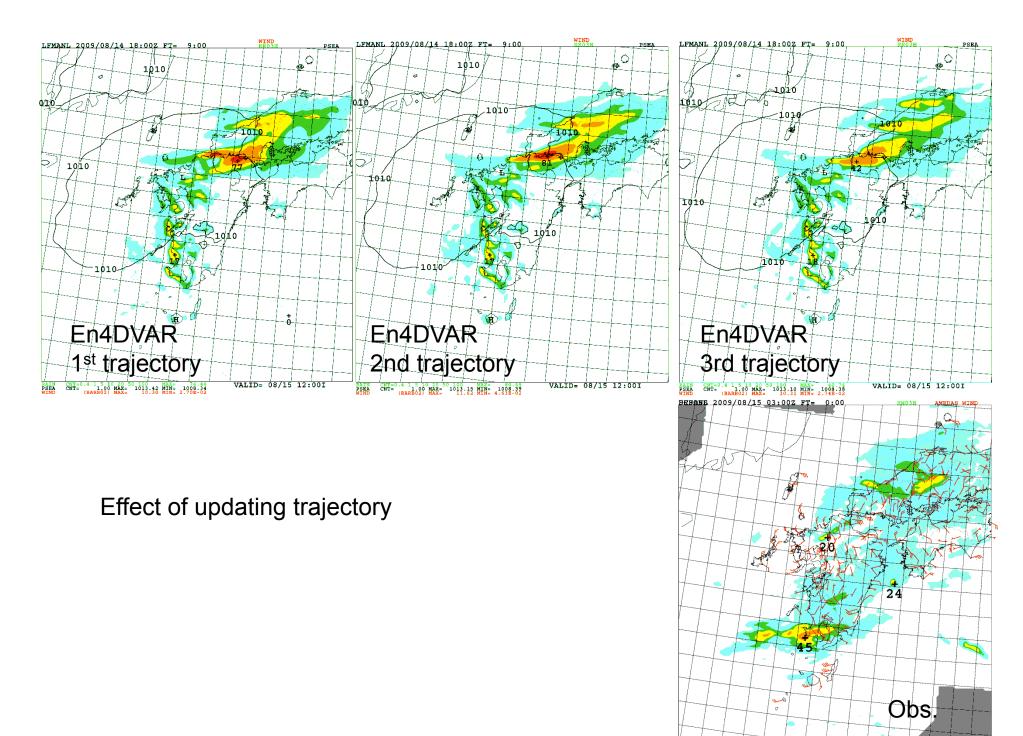




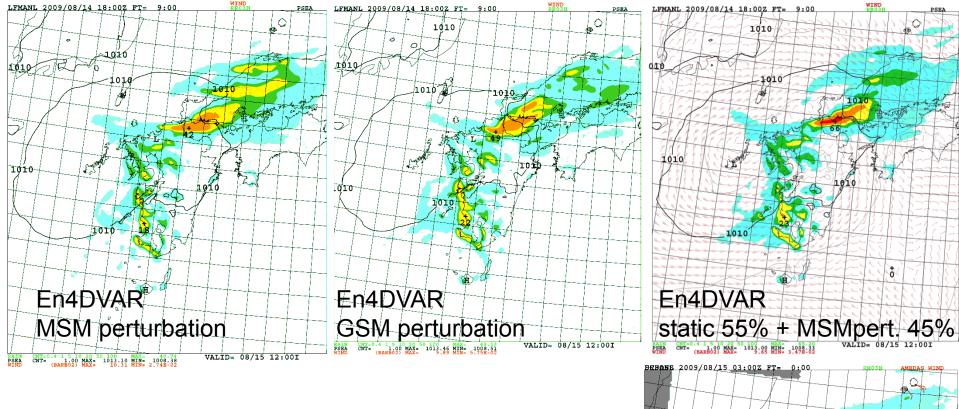
3DVAR

En3DVAR Static 55%, 4 scales Operational (+larger domain +prec., Ps Obs -low resolution inner)



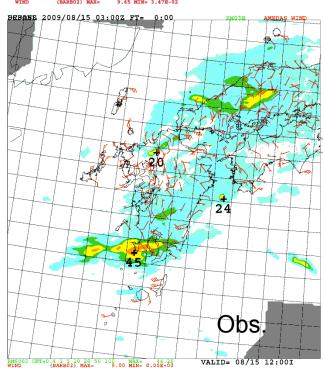


HG0003 CNT=0 4 1 5 10 20 50 100 WIND (BARBOZ) MAX= 9.00 MIN= 0.002+00 VALID= 08/15 12:001



Using different perturbations

MSM: operational mesoscale model (JMANHM) GSM: operational global model



summary

Perturbations with information of uncertainties in different scales are tried in variational method.

3d and 4d versions are implemented.

Independent control variable is assigned to each scale.

Localization scale is specified according to the displacement scale of each perturbation.

Perturbations are designed to represent displacement error.

Low cost method to represent flow-dependent background error with small ensemble size.

Possibility of construct an ensemble with En4DVARs with different perturbation design. (different trajectories, different weights on static B, etc.)