# Ensemble data assimilation at Japanese Met. Agency

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# Role of data assimilation



**Data Assimilation** 

#### Initialize model predictions; essential in NWP



# Data assimilation in NWP



time

#### Data Assimilation = Analysis



## Probabilistic view



## A schematic of EnKF



### A core concept of EnKF

Complementary relationship between data assimilation and ensemble forecasting



ANL Error

Ensemble Forecasting

This cycle process = EnKF

Analyze with the flow-dependent forecast error, ensemble forecast with initial ensemble reflecting the analysis error

#### Difference between EnKF and 3D-Var



#### An example of EnKF analysis accuracy

EnKF is advantageous to traditional data assimilation methods including 3D-Var, currently in operations at several NWP centers.



#### EnKF - summary

- EnKF considers flow-dependent error structures, or the "errors of the day"
  - "advanced" data assimilation method
    - 4D-Var is also an "advanced" method. How different?
- EnKF analyzes the analysis errors in addition to analysis itself

- "ideal" ensemble perturbations

# EnKF vs. 4D-Var

	EnKF	4D-Var		
"advanced" method?	Y	Y		
Simple to code?	Y	N (e.g., Minimizer)		
Adjoint model?	Ν	Y		
Observation operator	Only forward	Adjoint required		
	(e.g., TC center)			
Asynchronous obs?	Y (4D-EnKF)	Y (intrinsic)		
Analysis errors?	Y (ensemble ptb)	Ν		
Limitation	ensemble size	Assim. window		
	EnKF with infinite	EnKF with infinite ensemble size and 4D-		
	Var with infinite window are equivalent.			

#### LETKF (Hunt 2005; Hunt et al. 2007; Ott et al. 2004)

• Two categories of the EnKF (Ensemble Kalman Filter)

Perturbed observation	Square root filter (SRF)
(PO) method	
Classical	Relatively new
Already in operations (Canadian EPS)	Not in operations yet
Additional sampling errors by PO	No such additional sampling errors

- LETKF (Local Ensemble Transform Kalman Filter)
  - is a kind of ensemble square root filter (SRF)
  - is efficient with the parallel architecture

## KF and EnKF

Kalman Filter	Ensemble Kalman Filter
$\mathbf{x}:[N] \qquad \mathbf{P}:[N \times N]$	$\mathbf{X}:[N \times m]$
Forecast equations	<u>Ensemble forecasts</u>
$\mathbf{x}_{i}^{f} = M(\mathbf{x}_{i-1}^{a})$	$\mathbf{X}_{i}^{f} = [M(\mathbf{x}_{i-1}^{a(1)})   \cdots   M(\mathbf{x}_{i-1}^{a(m)})]$
$\mathbf{P}_{i}^{f} = \mathbf{M}_{\mathbf{x}_{i-1}^{a}} \mathbf{P}_{i-1}^{a} \mathbf{M}_{\mathbf{x}_{i-1}^{a}}^{T} + \mathbf{Q}$	$\equiv M(\mathbf{X}_{i-1}^{a})$ Approximated by $\mathbf{P}_{i}^{f} \approx \frac{\delta \mathbf{X}_{i}^{f} (\delta \mathbf{X}_{i}^{f})^{T}}{\mathbf{P}_{i}^{f}}$
<b>T</b> Z 1 ·	m-1
<u>Kalman gain</u>	$\partial \mathbf{Y} = H(\partial \mathbf{X}) : [p \times m]$
$\mathbf{K}_{i} = \mathbf{P}_{i}^{f} \mathbf{H}^{T} [\mathbf{H} \mathbf{P}_{i}^{f} \mathbf{H}^{T} + \mathbf{R}]^{-1}$	$\mathbf{K}_{i} = \boldsymbol{\delta} \mathbf{X}_{i}^{f} (\boldsymbol{\delta} \mathbf{Y})^{T} [\boldsymbol{\delta} \mathbf{Y} (\boldsymbol{\delta} \mathbf{Y})^{T} + (m-1)\mathbf{R}]^{-1}$
	[pxp] matrix inverse
Analysis equations	
$\mathbf{x}_i^a = \mathbf{x}_i^f + \mathbf{K}_i(\mathbf{y}_i^o - H(\mathbf{x}_i^f)) - $	$\rightarrow$ Solve for the ensemble mean
$\mathbf{P}^a = [\mathbf{I} - \mathbf{K} \ \mathbf{H}] \mathbf{P}^f$	→Ensemble perturbations
	$\delta \mathbf{X}^a = \left[ (m-1)\mathbf{P}^a \right]^{1/2}$

#### LETKF algorithm (Hunt, 2005, et al., 2007)

$$\mathbf{P}^{f} \approx \frac{\delta \mathbf{X}^{f} (\delta \mathbf{X}^{f})^{T}}{m-1} = \delta \mathbf{X}^{f} \widetilde{\mathbf{P}}^{f} (\delta \mathbf{X}^{f})^{T} \qquad \widetilde{\mathbf{P}}^{f} = \frac{\mathbf{I}}{m-1} : [m \times m]$$
  
In the space spanned by  $\delta \mathbf{X}^{f}$ 

$$\widetilde{\mathbf{P}}^{a} = [(\underline{m-1})\mathbf{I} / \rho + (\delta \mathbf{Y})^{T} \mathbf{R}^{-1} \delta \mathbf{Y}]^{-1} = \mathbf{U}\mathbf{D}^{-1}\mathbf{U}^{T}$$
  
Eigenvalue decomposition:  $\mathbf{U}\mathbf{D}\mathbf{U}^{T} : [m \times m]$ 

Analysis equations

$$\overline{\mathbf{x}}^{a} = \overline{\mathbf{x}}^{f} + \delta \mathbf{X}^{f} \widetilde{\mathbf{P}}^{a} (\delta \mathbf{Y})^{T} \mathbf{R}^{-1} (\mathbf{y}^{o} - H(\mathbf{x}^{f}))$$
$$\delta \mathbf{X}^{a} = \delta \mathbf{X}^{f} [(m-1)\widetilde{\mathbf{P}}^{a}]^{1/2} = \delta \mathbf{X}^{f} \sqrt{m-1} \mathbf{U} \mathbf{D}^{-1/2} \mathbf{U}^{T}$$

 $\frac{LETKF \text{ analysis}}{\mathbf{X}^{a} = \overline{\mathbf{x}}^{f} + \underbrace{\delta \mathbf{X}^{f} \left( \widetilde{\mathbf{P}}^{a} (\delta \mathbf{Y})^{T} \mathbf{R}^{-1} (\mathbf{y}^{o} - \overline{H(\mathbf{x}^{f})}) + \sqrt{m - 1} \mathbf{U} \mathbf{D}^{-1/2} \mathbf{U}^{T} \right)}_{\text{Ensemble analysis increments}}$ 

## **Research** activities

#### **MY MISSION:**

Developing a next-generation data assimilation system to improve operational NWP at JMA

#### Path to operations

- Develop and test LETKF (Hunt 2005; Hunt et al. 2007; Ott 1. et al. 2004; Hunt et al. 2004) with the Earth Simulator
- Develop LETKF with the JMA nonhydrostatic model 2.
- 3. Develop LETKF with the JMA global model
- Assess LETKF under the quasi-operational setup 4.

#### <u>Researches using the Earth Simulator LETKF system</u>

- Experimental 1.5-yr reanalysis (ALERA)
- Collaborative work with observing scientists
  LETKF with the Earth Simulator coupled atmos-ocean GCM

# Outline

- Developments of LETKF toward operations
  - Recent improvements
  - Quasi-operational comparison with 4D-Var
  - Probabilistic forecast skills
- Research projects with the Earth Simulator
  - Experimental Ensemble Reanalysis: ALERA
  - Collaboration with observing scientists

#### Developments toward operations

## LETKF developments at JMA

- LETKF (Local Ensemble Transform Kalman Filter, U of MD, Hunt et al. 2007; Ott et al. 2004) has been applied to 3 models
  - AFES (AGCM for the Earth Simulator)
     Miyoshi and Yamane, 2007: *Mon. Wea. Rev.*, 3841-3861.
     Miyoshi, Yamane, and Enomoto, 2007: *SOLA*, 45-48.
  - NHM (JMA nonhydrostatic model) Miyoshi and Aranami, 2006: SOLA, 128-131.
  - GSM (JMA global spectral model) Miyoshi and Sato, 2007: *SOLA*, 37-40.

#### Analysis-Forecast cycle experiment



#### Quasi-operational Experimental System



## Recent improvements

- Assimilation of satellite radiances
  - greatly improves the analysis accuracy Miyoshi and Sato, 2007: SOLA, 37-40.
- Removing local patches
  - solves the discontinuity problem near the Poles Miyoshi et al., 2007: SOLA, 89-92.
- Efficient MPI parallel implementation
  - solves the load imbalance problem
  - accelerates by a factor of 3
    - about 30% faster than operational 4D-Var with similar settings
- Adaptive satellite bias correction
  - a new idea analogous to the variational bias correction
  - showing great positive impact

# LETKF without local patches

SLP SPREAD (hPa) w/ LOCAL PATCH



SLP analysis ensemble spread after the first analysis step

The discontinuities caused by the local patches disappear.



Miyoshi et al. 2007

## Efficient parallel implementation



In the case of 9 comp. nodes

Irregular observing network causes significant load imbalances

Revising the node separation, we solved the load-imbalance problem almost completely; ~3 times faster computation

#### Impact by satellite radiances



RMSE and bias against radiosondes Blue: w/o satellite radiances Red: w/ satellite radiances

Reduced negative bias of Z and T

Reduced RMSE of Z in midupper troposphere (500-100hPa), especially in the SH and Tropics

#### Typhoon Rananim, August 2004



#### TC track ensemble prediction



### Statistical typhoon track errors



#### However, there was a problem



## Satellite radiance bias correction

Observation y has a bias b

 $b = b^{scan} + b^{air}$ Air mass bias (dependent on atmospheric state) Scan bias (constant) Statistically estimated offline Coefficients  $\beta$  of predictors p are estimated statistically  $b^{air} = p^T \beta$  
 Zenith angle

 Surface temperature

 Constant

 etc.

## Adaptive bias correction

Coefficients would change partly due to the deterioration of sensors

Allow temporal variation of the coefficients using data assimilation

Variational bias correction (e.g., Dee 2003; Sato 2007)

$$J(x) = \frac{1}{2}(x - x^{f})B^{-1}(x - x^{f})^{T} + \frac{1}{2}(y - Hx^{f})R^{-1}(y - Hx^{f})^{T}$$

$$J(x,\beta) = \frac{1}{2}(x-x^{f})B^{-1}(x-x^{f})^{T} + \frac{1}{2}(\beta-\beta^{f})B_{\beta}^{-1}(\beta-\beta^{f})^{T}$$

$$+\frac{1}{2}(y - p^T \beta - Hx^f)R^{-1}(y - p^T \beta - Hx^f)^T$$

Find the minimizer  $\beta$  of the cost function J through the variational procedure

#### Adaptive bias correction with LETKF

Analytical solution of the variational problem: minimizer (x,  $\beta$ )  $\begin{cases} \delta x = (B_x^{-1} + H^T R^{-1} H)^{-1} H^T R^{-1} (d - p^T \delta \beta) \\ \delta \beta = (B_\beta^{-1} + p R^{-1} p^T)^{-1} p R^{-1} (d - H \delta x) \end{cases}$ 

Adaptive bias correction with LETKF

- 1. Solve the LETKF data assimilation problem first  $\delta x = B_x H^T (HB_x H^T + R)^{-1} d = (B_x^{-1} + H^T R^{-1} H)^{-1} H^T R^{-1} d$   $-p^T \delta \beta \text{ difference}$
- 2. Solve the equation for  $\beta$  explicitly  $\delta \beta = (B_{\beta}^{-1} + pR^{-1}p^{T})^{-1}pR^{-1}(d - H\delta x)$ This coincides with the variational BC

#### Time series of bias coefficients



AMSU-A 4ch (sensitive to middle-lower troposphere) indicates significant drift from those estimated by 4D-Var

AMSU-A 6ch (sensitive to upper troposhere) and other sensors/channels indicate no significant drift

## Impact by adaptive bias correction



# Bias reduction



#### Improvement (%) relative to 4D-Var

	PseaSurf	T850	Z500	Wspd850	Wspd250
Global	-9.00	_10.45	-10.64	2.38	0.13
N. Hem.	-4.47	-2.95	-1.72	3.74	0.66
Tropics	0.48	_11.66	_17.60	11.69	9.88
S. Hem.	-10.90	-14.51	-13.00	-1.52	-3.81

#### Apply adaptive bias correction

	PseaSurf	T850	Z500	Wspd850	Wspd250
Global	-6.19	-4.36	-5.71	3.66	1.32
N. Hem.	-4.18	1.12	0.91	3.98	0.57
Tropics	6.86	3.39	3.09	14.07	10.21
S. Hem.	-7.60	-8.91	-7.91	-0.08	-1.62

#### Some bugs fixed in surface emissivity calculation

	PseaSurf	T850	Z500	Wspd850	Wspd250
Global	-5.21	-2.33	-4.21	3.94	1.73
N. Hem.	-3.89	2.06	1.32	4.30	1.30
Tropics	7.05	6.49	7.44	13.58	9.57
S. Hem.	-6.35	-6.47	-6.20	0.39	-1.14

Period: August 2004

Comparison with 3D-Var



Period: August 2004
#### Reason for bias drifts in AMSU-A 4ch

#### FACTS:

- ✓ 4D-Var uses RTTOV-7
- ✓ LETKF uses RTTOV-8

✓ AMSU-A ch.4 is sensitive to surface emissivity and lower tropospheric temperature

✓ A known bug in the surface emissivity model "FASTEM-2" in RTTOV-7, where the surface emissivity is spuriously overestimated

- 4D-Var VarBC corrects the "spurious" bias caused by the bug
- Therefore, observed radiances (bias corrected) are too large for LETKF
- Thus, the lower troposphere is heated by assimilating the too large radiance observations, which explains the cold forecast bias relative to analysis (because analysis is too warm)
- The adaptive BC within LETKF corrects the wrong bias

#### Experiments without satellite radiances



## Computational time

LETKF	4D-Var	
11 min x 60 nodes	17 min x 60 nodes	
5 min for LETKF		
6 min for 9-hr ensemble forecasts		
TL319/L60/M50	Inner: T159/L60	
	Outer: TL959/L60	

Estimated for a proposed next generation operational condition

6 min (measured) x 8 nodes for LETKF with TL159/L40/M50

Computation of LETKF is reasonably fast, good for the operational use.

#### Probabilistic forecast skills





#### Probabilistic forecast skills



# Summary

• Comparison with 3/4D-Var:

NH	LETKF ~ or > 4D-Var >> 3D-Var
Tropics	LETKF >> 4D-Var >> 3D-Var
SH	4D-Var >> LETKF > 3D-Var

- LETKF is advantageous in Typhoon prediction
- Probabilistic forecast skill is generally improved
  Still a bias problem exists
- Need improvements in the treatment of satellite radiances

### Future plan, ongoing development

- Better use of satellite radiances
  - QC system with RTTOV-8
    - Collaborating with satellite scientists
- Adapting to a higher resolution
  - Next-generation global model with the reduced
    Gaussian grid at a TL319/L60 resolution
  - kd-tree search algorithm to select local obs (geographical range search) will be implemented cf. Eric Kostelich already applied kd-tree at UMD

Make comparisons with the next-generation operational system

### Researches with the Earth Simulator

### Ensemble Reanalysis: ALERA

T159/L48 AFES (AGCM for the ES) LETKF w/ 40 members Assimilate real observations used in the JMA operational global analysis, except for satellite radiances Reanalysis from May 2005 through February 2007 ALERA

(AFES-LETKF Experimental Ensemble Reanalysis)

#### ALERA dataset

# ALERA

#### (AFES-LETKF Experimental Ensemble Reanalysis)

data are now available online for free!!

#### http://www3.es.jamstec.go.jp/

Contents

Ensemble reanalysis dataset for over 1.5 years since May 1, 2005

≻40 ensemble members

- ➢ensemble mean
- ➢ensemble spread

Available 'AS-IS' for free ONLY for research purposes Any feedback is greatly appreciated.

### Stable performance



### Ensemble spread and RMS diff



- Generally similar pattern
- Spread may need to be calibrated
  - (Spread should be smaller since NNR contains errors)
- > Underestimated spread by dense observations

## Analyzing the analysis errors

- EnKF provides not only analysis itself but also the analysis errors (or uncertainties of the analysis)
- What is the dynamical meaning of the analysis errors?



#### QBO and ensemble spread



## Stratospheric sudden warming

#### ALERA



#### Large spread in tropical lower wind

ALERA



Courtesy of T. Enomoto

### Tropical lower wind and the spread



# Lag correlation between mean and spread



Courtesy of T. Enomoto

#### Collaboration with observing scientists

There was an intensive observing project in the tropical western Pacific (a.k.a. PALAU-2005); the dropsonde obs during June 12-17, 2005 have been assimilated with the AFES-LETKF system.



Satellite image and dropsonde locations

Moteki et al., 2007: SOLA

#### Collaboration with observing scientists

#### Impact by dropsonde observations



Moteki et al., 2007: SOLA

#### Collaboration with observing scientists



#### Propagation of observing signals

cf. Szunyogh et al. (2000; 2002) Hodyss and Majumdar (2007)

Faster propagation than the advection speed, the faster speed (~12 m/s) corresponds to Rossby wave propagation

Moteki et al., 2007: SOLA

## Summary

- Ensemble spread represents errors well.
- There seems to be dynamical meanings of analysis ensemble spread, which could be investigated in various scales.
- Collaborative study with observing scientists has just begun.

#### Research plans with the ES

- ALERA-2
  - 5-yr reanalysis (21<sup>st</sup> century reanalysis)
  - AFES-MATSIRO (SiB) (atmos-land coupled)
  - More diagnostics (OLR, Precipitation, land variables, etc.; any requests?)
- CFES-LETKF
  - LETKF with coupled atmos-land-ocean model
- More observing projects (e.g., PALAU-2008)

#### Other ongoing activities with LETKF

- Collaboration with chemical/aerosol-transport modeling scientists
  - Dr. Thomas Sekiyama at the Met Research Institute, JMA
  - Dr. Nich Schutgens at the Center for Climate System Research, University of Tokyo
- Collaboration with university students and scientists to assimilate atmospheric lidar wind data into a nonhydrostatic fine-mesh model

- Prof. Toshiki Iwasaki at the Tohoku University

### Collaborators

#### <u>AFES-LETKF</u>

- Prof. Shozo Yamane (Chiba Institute of Science and FRCGC/JAMSTEC, AFES)
- Dr. Takeshi Enomoto (ESC/JAMSTEC, AFES)
- Dr. Qoosaku Moteki (IORGC/JAMSTEC, Observing scientist)

#### <u>NHM-LETKF</u>

- Kohei Aranami (NPD/JMA, NHM)
- Dr. Hiromu Seko (MRI/JMA, DA with NHM)

#### <u>GSM-LETKF</u>

- Yoshiaki Sato (NPD/JMA, staying at NCEP)
- Takashi Kadowaki (NPD/JMA, 4D-Var)
- Ryota Sakai (NPD/JMA, EPS)
- Munehiko Yamaguchi (NPD/JMA, Typhoon EPS)

#### Chemical Transport

– Dr. Thomas Sekiyama (MRI/JMA, Chemical model)

### Acknowledgments

- The idea of LETKF has been developed at the University of Maryland.
- Radiative transfer model RTTOV-8 is provided by **EUMETSAT**.
- ALERA has been produced using the Earth Simulator.



#### Supplement: ALERA accuracy

# Snapshot (SLP)



#### ALERA analysis is almost identical to NNR.

#### Zonal mean winds in JJA



#### Comparison with observations



### Compared with radiosondes



#### Forecast RMS errors



#### Supplement: GSM-LETKF

#### Improvement (%) relative to 4D-Var

#### August 2004

	PseaSurf	T850	Z500	Wspd850	Wspd250
Global	-2.53	-2.04	-1.57	5.51	3.56
N. Hem.	-2.77	2.16	2.09	4.91	1.63
Tropics	7.05	4.90	4.95	14.92	11.81
S. Hem.	-3.06	-5.49	-2.68	2.62	1.91

#### December 2005

	PseaSurf	T850	Z500	Wspd850	Wspd250
Global	-3.39	-1.27	-2.60	4.91	2.69
N. Hem.	-5.60	-0.29	-4.46	1.16	0.83
Tropics	8.74	4.94	13.43	16.52	12.10
S. Hem.	-1.06	-4.98	-0.82	3.79	0.15

#### LETKF is advantageous in the summer hemisphere

High T >  $T_{CLM}$  + 2K T850 gt2K NH ΙΕΤΚΕ 0.95 B٧ 0.9 **ROC Area** 0.85 0.8 0.75


Low  $T < T_{CLM} - 2K$ T850 lt-2K NH LETKF 0.95 ΒV 0.9 **ROC Area** 0.85 0.8 0.75 0.7 24 48 72 96 120 144 168 192 216 0 Forecast Time [hr]

## Ensemble size $20 \rightarrow 50$



## RMSE and bias against radiosondes

Blue: Operational 4D-Var Red: 20-member LETKF Green: 50-member LETKF

50 members > 20 members

Generally 4D-Var > LETKF Exception: mid-upper tropospheric temperature in the SH

w/ satellite radiances