

The CNMCA Operational LETKF Implementation: Description and Results

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and contributions of

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Outline

- The CNMCA Ensemble Data Assimilation System:
 - Local Ensemble Transform Kalman Filter (LETKF)
 - Background Covariance Inflaction
- Control State LETKF
- Comparison with CNMCA 3DVar and IFS 4DVar analysis
- Tests with AMSU-A radiances and Outer Loop
- Future developments



Data Assimilation at CNMCA

The first data assimilation cycle CNMCA was based on an OI scheme, then a 3DVar algorithm was implemented. In 2006 the following questions were raised at CNMCA:

1.Can we improve on our currently operational 3DVar without the complications of 4DVar?

2. Can we develop a system to consistently evaluate initial and forecast uncertainties?

Ensemble data assimilation (EnDA) has proved to be a viable and competitive alternative to variational methods (*Houtekamer*, 2005; *Fertig et al.*, 2007; *Miyoshi*, 2007)

Ensemble data assimilation provides "optimal" analysis errors estimates for ensemble forecasting and it overcomes the need for ad hoc inverse methods





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Ensemble Kalman Filter DA

• At CNMCA the **LETKF** (Hunt et al. 2007) formulation was chosen, because algorithmically simple to code, intrinsically parallel, etc.

• The analysis is done in the space of the ensemble perturbations and computed separately at each grid point selecting only the obs in a vicinity. This explicit localization reduces the problem dimensionality and the spurious correlations between distant locations due to limited ensemble size

Analysis Ensemble Mean	$\overline{\mathbf{x}}^{\mathbf{a}} = \overline{\mathbf{x}}^{\mathbf{b}} + \mathbf{X}^{\mathbf{b}} \overline{\mathbf{w}}^{\mathbf{a}}$
Analysis Ensemble Perturb.	$\mathbf{X}^{a} = \mathbf{X}^{b} \mathbf{W}^{a}$
Analysis Ensemble	$\mathbf{x}^{\mathbf{a}} = \mathbf{x}^{\mathbf{b}} + \mathbf{X}^{\mathbf{b}} \mathbf{w}^{\mathbf{a}}$

$$\overline{W}^{a} = \widetilde{P}^{a}Y^{bT}R^{-1}(y - H(\overline{x}^{b}))$$

$$\widetilde{P}^{a} = [(m-1)I + Y^{bT}R^{-1}Y^{b}]^{-1}$$

$$Y^{b} = [(H(x_{1}^{b}) - \overline{H(x^{b})}, \dots, (H(x_{m}^{b}) - \overline{H(x^{b})})]$$

$$W^{a} = [(m-1)\widetilde{P}^{a}]$$

$$w^{a} = W^{a} + [\overline{W}^{a}, \dots, \overline{W}^{a}]$$

 ensemble mean analysis is the linear combination of forecast ensemble states which best fits the observational dataset
 analysis ensemble members are locally linear combinations of background ensemble members



CNMCA NWP SYSTEM since 1 June 11

Ensemble Data Assimilation:



LETKF analysis ensemble (40 members) every 6h using TEMP, PILOT, SYNOP, SHIP, BUOY, Wind Profiler, AMDAR-ACAR-AIREP, MSG AMV, METOP/ERS2 scatt. winds, NOAA/METOP AMSUA radiances (very soon) + Land SAF snow mask,

2.8 km

50 v.l.

10°E

٥٥

(2.8Km)

10°E

- compressible equations - explicit convection

30°E

20°E

30°E

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Covariance Inflaction

In the CNMCA LETKF implementation, model errors and sampling errors are taken into account using:

- State Dependent Multiplicative Inflaction according to Whitaker et al (2010)

an. pert.
$$\mathbf{x}'_{a} = \mathbf{x}'_{a} \sqrt{\alpha \frac{\sigma_{b}^{2} - \sigma_{a}^{2}}{\sigma_{a}^{2}} + 1}}$$
 $\alpha = 0.95$
 $\sigma^{2} = variance$

- Climatological Additive Noise

an. memb. $\mathbf{x}_i^a \leftarrow \mathbf{x}_i^a + lpha \mathbf{x}_i^n$, $lpha \mathbf{x}_i^n \sim N(\mathbf{0}, \mathbf{Q})$

 \mathbf{X}_{i}^{n} randomly selected, 48-24h forecast differences

- Lateral Boundary Condition Perturbation using EPS
- Climatological Perturbed SST





Ensemble Background Spread



20 August 2011 00UTC: 200 hPa Temperature Spread

Ensemble Background Spread



20 - 26 August 2011 00UTC: 200 hPa Temperature Spread

Ensemble Background Spread



26 August 2011 00UTC: 200 hPa Temperature Spread









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3DVar/LETKF comparison WIND (m/s)00 UTC FC + 12 h



3DVar/LETKF comparison

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3DVar/LETKF comparison

NMC

Italian stations April and May 2011





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Thursday 23 June 2011 00UTC ROME Forecast t+6 VT: Thursday 23 June 2011 06UTC Surface: total precipitation COSMO_ME: precipitation in the previous 06 hour interval in a 06 hours interval







Control/Mean LETKF comparison



NMC

Comparison of LETKF/IFS analysis

The deterministic COSMO-ME runs, initialized by the 00 UTC LETKF analysis ensemble mean and by the global IFS 4DVAR system and driven by IFS boundary conditions, are objectively verified against the European observation network. In this evaluation you have to take into account the reduced number of observation types (no radiances in this experiment) used in the CNMCA LETKF system.



IFS/LETKF analysis comparison







Radiances Treatment

AMSU-A Assimilation

Selecting which satellite radiances to assimilate is complicated by the fact that they not have a single well-defined vertical location The weighting function at a particular model point indicates the sensitivity of that observation to the state at that model grid point \rightarrow "MAXIMUM-BASED SELECTION" METHOD

(Fertig et al. 2007)

AMSU-A are treated as "single-level" observations

Assign radiance observations to the model level for which the magnitude of the weighting function (wf) is largest.

 Use also the wf shape as vertical covariance localization function



p(hPa)

Weighting function (transmittance vert. derivative)

 $w_k = (\tau_{v,k-1} - \tau_{v,k}) / (\ln(p_k) - \ln(p_k))$

Impact of AMSUA rad.

600

800

1000

-1

Relative difference (%) in RMSE computed against IFS analysis for 00 UTC EuroHRM runs (28km) from 11-10-2009 to 10-11-2009

RTTOV 9.3



step 36 Temperature





-0.5



0.5

0



Handling Non-Linearities Results

No-cost LETKF smoother works:

LETKF an. $\overline{x}_{n}^{a} = \overline{x}_{n}^{b} + X_{n}^{b} \overline{w}_{n}^{a}$ Smoothed an. $\overline{x}_{n-1}^{a} = \overline{x}_{n-1}^{a} + X_{n-1}^{a} \overline{w}_{n}^{a}$

Relative difference (%) in RMSE computed against IFS analysis step 36 Wind Vector step 48 Wind Vector 200 200 400 400 600 600 800 800 1000 1000 ñ. ñ -1

 We tested Quasi Outer Loop and Running in Place algorithms (Yang and Kalnay, 2010) without any success

• We have not be able to have improvements from other variants of Outer Loop





Conclusions and Future Developments

As far as we know, CNMCA is the first meteorological centre which uses operationally a pure ensemble data assimilation (LETKF) to initialize a deterministic NWP model (COSMO-ME).
The long run is integrated from a control state LETKF derived using a control forecast instead of the FG ensemble mean

- The use of AMSU-A has improved the LETKF analysis.
- Assimilation of AMSU-B/MHS and IASI retrievals will be investigated soon.
- Balancing and non-linearities are issues to address
- Tests with COSMO model and shorter assimilation window
- Further tuning of model error representation (tuning of cov. localization, evolved additive noise, bias correction, etc.)
- Implement a Short-Range EPS based on LETKF





Thanks for your attention!



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