

Weakly Constrained LETKF for Convective-Scale Data Assimilation

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Our study

- ▶ Data assimilation (DA) on convective scales should update hydrometeors
- ▶ Assimilation of observations such as radar reflectivity or cloud products are important for prediction on convective scales
- ▶ [Janjic et al. 2014](#) show that is important to preserve both positivity and mass with DA algorithm when estimating variables that should be non-negative
- ▶ Operationally used DA algorithms when updating hydrometeors clip negative values to zeros, modifying mass
- ▶ Here, we propose a fast, easy to implement modification of LETKF that is able to [weakly](#) preserve both properties of mass conservation for each hydrometeor variable and non-negativity.

Weakly Constrained LETKF

$$\begin{aligned}\min_{\mathbf{x}} J(\mathbf{x}) &= J_b(\mathbf{x}) + J_o(\mathbf{x}) + J_m(\mathbf{x}) \\ &= \frac{1}{2} (\bar{\mathbf{x}}_k^b - \mathbf{x}) \mathbf{P}_k^{b^{-1}} (\bar{\mathbf{x}}_k^b - \mathbf{x})^T + \frac{1}{2} [\mathbf{y}_k^o - \mathbf{H}(\mathbf{x})] \mathbf{R}_k^{-1} [\mathbf{y}_k^o - \mathbf{H}(\mathbf{x})]^T \\ &\quad + \frac{1}{2} [\mathbf{m}_k - \mathbf{S}(\mathbf{x})] \mathbf{M}_k^{-1} [\mathbf{m}_k - \mathbf{S}(\mathbf{x})]^T\end{aligned}$$

- ▶ \mathbf{m} is a vector quantity, whose elements are the domainwise (global) integral of hydrometeors.
- ▶ \mathbf{S} is operator which calculates the domainwise (global) integral for each of the microphysical species,

Constraint on mass is up to accuracy \mathbf{M}_k

$$\mathbf{M}_k = \frac{1}{N_{ens} - 1} \sum_{i=1}^{N_{ens}} [\mathbf{m}_k^* - \mathbf{S}(\mathbf{x}_k^{b(i)})] [\mathbf{m}_k^* - \mathbf{S}(\mathbf{x}_k^{b(i)})]^T$$

Weakly Constrained LETKF

For mass:

$$\bar{\mathbf{x}}_k^{a,M} = \bar{\mathbf{x}}_k^b + \mathbf{P}_k^{a,M} \mathbf{H}^T \mathbf{R}_k^{-1} (\mathbf{y}_k^o - \bar{\mathbf{y}}_k^b) + \frac{1}{N_{ens} - 1} \mathbf{x}_k^{a,M} (\mathbf{S} \mathbf{x}_k^{a,M})^T \mathbf{M}_k^{-1} [\mathbf{m}_k - \mathbf{S} (\bar{\mathbf{x}}_k^b)]^T.$$

$$\mathbf{W}_k^{a,M} = \left[(N_{ens} - 1) \mathbf{I} + (\mathbf{H} \mathbf{x}_k^b)^T \mathbf{R}_k^{-1} \mathbf{H} \mathbf{x}_k^b + (\mathbf{S} \mathbf{x}_k^b)^T \mathbf{M}_k^{-1} \mathbf{S} \mathbf{x}_k^b \right]^{-1},$$

Weakly Constrained LETKF

For mass:

$$\bar{\mathbf{x}}_k^{a,M} = \bar{\mathbf{x}}_k^b + \mathbf{P}_k^{a,M} \mathbf{H}^T \mathbf{R}_k^{-1} (\mathbf{y}_k^o - \bar{\mathbf{y}}_k^b) + \frac{1}{N_{ens} - 1} \mathbf{x}_k^{a,M} (\mathbf{S} \mathbf{x}_k^{a,M})^T \mathbf{M}_k^{-1} [\mathbf{m}_k - \mathbf{S} (\bar{\mathbf{x}}_k^b)]^T.$$

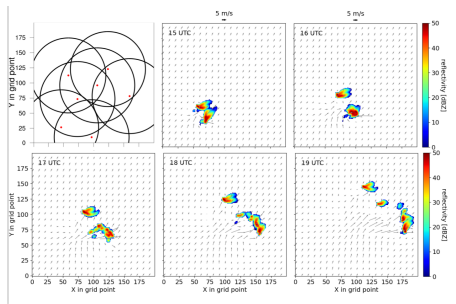
$$\mathbf{W}_k^{a,M} = \left[(N_{ens} - 1) \mathbf{I} + (\mathbf{H} \mathbf{x}_k^b)^T \mathbf{R}_k^{-1} \mathbf{H} \mathbf{x}_k^b + (\mathbf{S} \mathbf{x}_k^b)^T \mathbf{M}_k^{-1} \mathbf{S} \mathbf{x}_k^b \right]^{-1},$$

For positivity:

- ▶ To avoid spurious convection (Aksoy et al., 2009) and overestimated analysis increments (Zeng et al., 2021), a non-negative threshold value is set for very small reflectivities. We call those data **clear-air reflectivity data**
- ▶ Radar reflectivity data depend nonlinearly on hydrometeors. Further, reflectivity data (including clear-air reflectivity data) are available at radar observation locations, therefore not in every grid point of the model
- ▶ By assimilating additional clear-air reflectivity data, we are asking in the **approximate weak sense** that non-negativity is preserved in the analysis of hydrometeors

Experimental setup

Idealized setup for radar DA



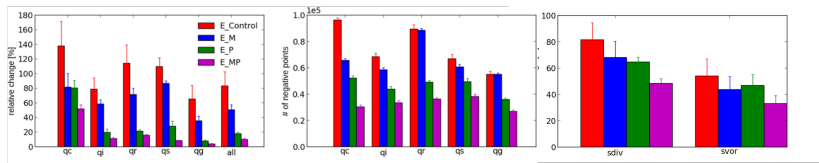
Zeng et al. 2021: Assimilating radar radial wind and reflectivity data in an idealized setup of the COSMO-KENDA system, *Atmospheric Research*, **249**, 105282, <https://doi.org/10.1016/j.atmosres.2020.105282>.

- ▶ COSMO model with a 2-km horizontal resolution
- ▶ Efficient Modular VOLUME scanning RADAR Operator (EMVORADO, Zeng et al., 2014, 2016)
- ▶ Both radial wind and reflectivity data are assimilated
- ▶ Ensemble size is 80
- ▶ Observations perturbed with Gaussian noise with a standard deviation of 5.0 dBZ and 1.0 m/s

Results: Impact of the different constraints

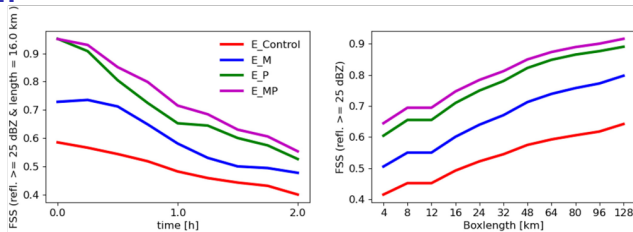
1. $E_{Control}$ Radar reflectivity and wind assimilated
2. E_M In addition mass constraint
3. E_P In addition positivity constraint (clear-air reflectivity data)
4. E_{MP} Both constraint

If clear-air reflectivity data are assimilated, a threshold value of 5 dBZ is set, that is, all reflectivity values smaller than 5 dBZ are set to 5 dBZ. If clear-air reflectivity data are not assimilated, all reflectivity values smaller than 5 dBZ are set to missing values.



Janjic, T. and Y. Zeng, 2021, Weakly constrained LETKF for estimation of hydrometeor variables in convective-scale data assimilation, *Geophysical Research Letters*, **48**, e2021GL094962, <https://doi.org/10.1029/2021GL094962>.

Conclusion



- ▶ Assimilation of clear-air reflectivity reduces the number of grid points with negative values of hydrometeor variables in pre-analyses and mitigates the biased increase and uncertainties in the total mass.
- ▶ Weak constraints on mass conservation is effective (especially in reducing mass bias) but not as effective as assimilating clear-air reflectivity.
- ▶ **The best results are obtained by a combination of both constraints.**
- ▶ Method proposed here is well suited for the assimilation of radar data in convection permitting models.
- ▶ It is appropriate for estimation of high dimensional state and requires only minor changes to the already existing implementation of LETKF.

Main References.



Janjic and Zeng 2021 " Weakly Constrained LETKF for Estimation of Hydrometeor Variables in Convective-Scale Data Assimilation", *Geophysical Research Letters*, **48**, e2021GL094962, <https://doi.org/10.1029/2021GL094962>



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Janjic, et al. 2014. "Conservation of mass and preservation of positivity with ensemble-type Kalman filter algorithms." *Monthly Weather Review* 142.2: 755-773.