



Quantify transport errors in CO₂ mixing ratios using a small-size ensemble

Thomas Lauvaux, Penn State University

1. Generate a calibrated ensemble which represents the characteristics of transport errors using a minimum number of members

PhD of Liza Diaz-Isaac (now at Scripps, UC San Diego)

2. Remove the sampling noise due to the limited number of members

in collaboration with Marc Bocquet (CEREA, Paris) and Nicolas Bousserez (U. of Colorado)

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Physics-based ensemble of atmospheric simulations



from Diaz-Isaac et al. (in prep.)

Description of the WRF simulation domain



Period: June 18-July 21, 2008 **Grid Resolution:**

Parent Domain (d01): 30-km Inner Domain (d02): 10-km

- Re-initialization every five days.
- A multi-physics and multi-analysis ensemble will be used instead of a randomly perturbed ensemble.
- The multi-physics and multi-analysis configurations allows us to select different model configurations for multiple atmospheric inversions.

from Diaz-Isaac et al. (in prep.)

Rank Histograms & Score



If the ensemble is under-dispersive *Than the score will be greater than 1*



How to improve the score of the ensemble, i.e. increase the dispersion?

Hamill, 2001

Calibration of the ensemble



Simulated Annealing 10-member Sub-Ensemble



from Diaz-Isaac et al. (in prep.)

Ensemble-based CO₂ Variance





PBL Height at DVN



from Diaz-Isaac et al. (in prep.)



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2. Remove the sampling noise due to the limited number of members

Work based on recent developments in NWP for noise filtering using an objective filter

Sub-question: How to define the reference (best ensemble)? Note: 45-member ensemble is not calibrated

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FIG. 1. Outline of the paper, from original theories to applications, via new optimality criteria definition.



from Menetrier et al. (2015), Monthly Weather Review

Impact of the calibration on Filtered CO₂ Variances and Covariances using calibrated and random ensembles of 5 and 8members

> Monthly averages of the filtered hourly results 27 June to 20 July, 2008

1. "Random ensembles" were generated with randomly selected members

- Random ensembles ("random X-member") do not share any members with calibrated ensembles ("calibrated X-member") = members never selected by calibration procedure

- Random ensembles of different sizes (i.e. 5-member and 8-member) share only one member (I didn't do that on purpose, it just turned out to be that way)

2. "Calibrated ensembles" were generated with two methods:

- Simulated Annealing (SA): best option AND best option most similar to GA

- Genetic Algorithm (GA): best option AND best option most similar to SA

Note: before, I was only using "SA ensembles most similar to GA" (and not "best option")

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Optimal sub-ensemble = sufficient number of members (here 25 members) and calibration of the ensemble

Convergence of the filter for variances





Unfiltered variances averaged over one month (in ppm²)



Filtered variances averaged over one month (in ppm²)

An example

I show here an example of hourly variances for July 5 and July 20 of 2008



25-member variances averaged over one month (in ppm²)

And the error covariances

Here is the monthly averaged error correlations after using Menetrier's filter (Schur product here)

Filtering covariances (Schur product) of calibrated ensembles



Full (=45-member) ensemble error correlations averaged over one month

Filtering covariances (Schur product) of calibrated ensembles



Full (=45) and 25-member error correlations averaged over one month

Filtering covariances (Schur product) of calibrated ensembles



Conclusions and Perspectives

1. Generate a calibrated ensemble which represents the characteristics of transport errors using a minimum number of members

Starting from 45 members, small-size ensembles of 8 and 10 members are able to represent the observed CO₂ variances

Initial dispersion was small which limits our ability to match exactly the model errors

 \rightarrow Generate perturbed ensembles to increase the dispersion

2. Remove the sampling noise due to the limited number of members

Calibrated sub-ensembles show specific error structures not visible in the original ensemble and in random sub-ensembles

 \rightarrow Need to create error covariance matrices for the CO₂ atmospheric inversion