

On the guess consistency in multi-incremental multi-resolution variational data assimilation

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Abstract. Variational Data Assimilation (DA) schemes are often used to adress high dimensional non-linear problems in operational applications in the Numerical Wheather Prediction (NWP) domain. Because of the high computational cost of such minimization problems, various methods can be applied to improve the convergence at a reasonable numerical cost. One of these methods currently applied in operational DA schemes is the multi-incremental approach that consists in solving a succession of linearized versions of the original non-linear problem in several outer loops, by using well known algorithms to ensure the convergence of the linear problem at the inner loop level, and using the solution of the inner loops to update the problem at each outer loop. In order to save computational cost, the multi-incremental multi-resolution method consists in starting the minimization at a lower resolution than the original one, and increasing it at the outer loop level until the full resolution of the problem. In such a scheme, the way to compute the new guess at each outer loop from the previous iterations is crucial. We adress the question of the guess consistency in the standard method currently used in operational systems, and also present a new method which ensures the guess consistency and need simpler calculations.

1 Introduction

intro...

2 Data Assimilation Problem

15 presenting the general problem

2.1 Multi-incremental data assimilation

present the multi-incremental scheme.

2.2 Changing the resolution

in special cases, the background error covariance matrix can be updated between outer iterations defining \mathbf{B}_k and its square-root

20 \mathbf{U}_k . In this case, one can obtain the background increment as

$$\delta \bar{\mathbf{x}}_k^b = -\mathbf{B}_k^{-1} \sum_{i=1}^{k-1} \delta \mathbf{x}_i^a = -\sum_{i=1}^{k-1} \mathbf{B}_k^{-1} \mathbf{B}_i \delta \bar{\mathbf{x}}_i^a, \quad (1)$$

$$\delta \mathbf{v}_k^b = -\mathbf{U}_k^T \mathbf{B}_k^{-1} \sum_{i=1}^{k-1} \delta \mathbf{x}_i^a = -\sum_{i=1}^{k-1} \mathbf{U}_k^T \mathbf{B}_k^{-1} \mathbf{U}_i \delta \mathbf{v}_i^a. \quad (2)$$

If $\mathbf{B}_k^{-1} \mathbf{B}_i \delta \bar{\mathbf{x}}_i^a \neq \delta \bar{\mathbf{x}}_i^a$ or if $\mathbf{U}_k^T \mathbf{B}_k^{-1} \mathbf{U}_i \delta \mathbf{v}_i^a \neq \delta \mathbf{v}_i^a$, equation (??) cannot be used consistently. This is the case if the resolution
 25 increases between the outer loops for computational efficiency. Therefore, the \mathbf{B} matrix depends on k and one needs to use
 interpolators in model space $\mathbf{T}_{i \rightarrow k}^{\mathbf{x}} \in \mathbb{R}^{n_k \times n_i}$ and in control space $\mathbf{T}_{i \rightarrow k}^{\mathbf{v}} \in \mathbb{R}^{m_k \times m_i}$ from the resolution \mathcal{R}_i to \mathcal{R}_k , where n_k
 and m_k respectively denotes the size of the model space and the control space.

2.2.1 HEADING

TEXT

30 3 Conclusions

TEXT

Code availability. TEXT

Data availability. TEXT

Code and data availability. TEXT

35 *Sample availability.* TEXT

Video supplement. TEXT

Appendix A

A1

Author contributions. TEXT

40 *Competing interests.* TEXT

Disclaimer. TEXT

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References

REFERENCE 1

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