

# The University of Reading

## Approximate iterative methods for variational data assimilation

A.S. Lawless<sup>1</sup>, S. Gratton<sup>2</sup> and N.K. Nichols<sup>1</sup>

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<sup>1</sup>*Department of Mathematics  
The University of Reading  
Whiteknights, PO Box 220  
Reading  
Berkshire RG6 6AX*

<sup>2</sup>*CERFACS  
42 Avenue Gustave Coriolis  
31057 Toulouse CEDEX  
France*

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# Department of Mathematics





with  $x \in \mathbb{R}^n$  [3]. We can write the 4D-Var objective function (1) in this form by putting  $f(x) = C^{-1/2} \hat{d}$ , where

$$\hat{d}(x_0) = \begin{pmatrix} 0 \\ x_0 - x^b \\ H_0[x_0] - y_0^o \\ \vdots \\ H_n[x_n] - y_n^o \end{pmatrix}; \quad C^{-1} = \begin{pmatrix} B_0^{-1} & 0 \\ 0 & R^{-1} \end{pmatrix} \quad \text{with} \quad R = \text{diag}(R_i);$$

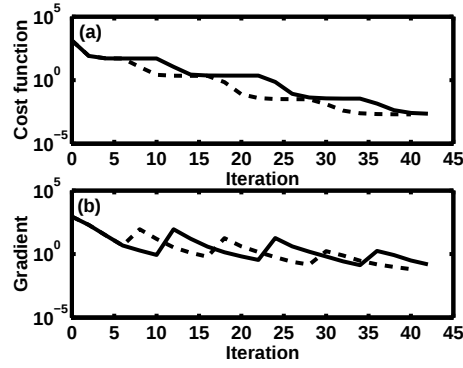


Figure 1: Comparison of convergence of (a) objective function and (b) its gradient for a constant convergence criterion (solid line) and a variable criterion (dashed line).

with  $D_t \psi = \partial_t \psi + u \partial_x \psi$ . In these equations  $\bar{h} = \bar{h}(x)$  is the height of the bottom orography,  $u$  is the velocity of the fluid and  $\bar{\phi} = gh$  is the geopotential, where  $g$  is the gravitational constant and  $h > 0$  the depth of the fluid above the orography. The problem is defined on the domain  $x \in [0; L]$ , with periodic boundary conditions, and we let  $t \in [0; T]$ .

The model is discretized using a two-time-level semi-implicit semi-Lagrangian scheme. Further details of the numerics can be found in [6]. We use a periodic domain of 1000 grid points, with a spacing  $\Delta x = 0.01$  m, so that  $x \in [0 \text{ m}; 10 \text{ m}]$ . The model time step is taken to be  $9.2 \times 10^{-3}$  s and we consider an assimilation over a window of 100 time steps.

In order to test the assimilation we run ‘identical twin’ experiments, in which observations are generated from a run of the model defined to be the truth. These observations are then assimilated using 4D-Var, starting from an incorrect prior estimate. The inner minimization is performed using a conjugate gradient method and is considered to



Again it is possible to provide a theoretical understanding of these results by an analysis of the (PGN) method. The method can be considered as a way of solving the approximate normal equations

$$\tilde{\mathbf{J}}(\tilde{\mathbf{x}}^*)^T \mathbf{f}(\tilde{\mathbf{x}}^*) = 0; \quad (14)$$

Then based on the work of [7] and [8] we can prove the following theorem:

**Theorem 2** *Let the first derivative of  $\tilde{\mathbf{J}}(\mathbf{x})^T \mathbf{f}(\tilde{\mathbf{x}})$  be written*

$$\tilde{\mathbf{J}}(\mathbf{x})^T \mathbf{J}(\mathbf{x}) + \tilde{\mathbf{Q}}(\mathbf{x}); \quad (15)$$

*where  $\tilde{\mathbf{Q}}(\mathbf{x})$  represents second order terms arising from the derivative of  $\tilde{\mathbf{J}}(\mathbf{x})$ . Suppose that on each iteration  $k$*

- [4] Ortega JM, Rheinboldt WC. *Iterative Solution of Nonlinear Equations in Several Variables*