Inverse modelling for trace gas surface flux estimation, impact of a nondiagonal B-matrix Ross Bannister, Univ of Reading

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One of the most appealing uses of data assimilation is to infer useful information about a dynamical system that is not observed directly. This is the case for the estimation of surface fluxes of trace gases (like methane). Such fluxes are not easy to measure directly on a global scale, but it is possible to measure the trace gas itself as it is transported around the globe. This is the purpose of INVICAT (the inverse modelling system of the chemical transport model TOMCAT), which has been developed here. INVICAT interprets observations of (e.g.) methane over a time window to estimate the initial conditions (ICs) and surface fluxes (SFs) of the TOMCAT model.

This poster will show how INVICAT has been expanded from a diagonal background error covariance matrix (B-matrix, DB) to allow an efficient representation of a non-diagonal B-matrix (NDB). The results of this process are mixed. A NDBmatrix for the SF field improves the analysis against independent data, but a NDB-matrix for the IC field sometimes appears to degrade the analysis. This poster presents these results and suggests that a possible reason for the degraded analyses is the presence of a possible bias in the system.

Why use a non-diagonal **B-matrix**?

Forward Model: TOMCAT Species of interest: CH₄ Number of grid points: 64 × 32 × 60 (5⁵/₈° × 5⁵/₈°) Period: First 100 days of 2018 **Chemistry:** OH, O(¹D), Cl $+ \mathbf{u} \cdot \nabla \mathbf{x}_t = \mathbf{x}_t^{\mathrm{sf}} + \mathbf{c}$ Prescribed Surface Chemistry winds CH₄ flux concentration **Inverse Model:** Control $\mathbf{x}_{t=0}^{ ext{ic}}$ $\mathbf{B} =$ vector: INVICAT $egin{aligned} \mathbf{x}_{t=0}^{ ext{sf}} \ \mathbf{x}_{t=1}^{ ext{sf}} \ \mathbf{x}_{t=2}^{ ext{sf}} \end{aligned}$ $J(\mathbf{x}) =$ $\frac{1}{2} \left(\mathbf{x} - \mathbf{x}^{b} \right)^{T} \mathbf{B}^{-1} \left(\mathbf{x} - \mathbf{x}^{b} \right) + \frac{1}{2} \left(\mathbf{y} - \mathbf{h}(\mathbf{x}) \right)^{T} \mathbf{R}^{-1} \left(\mathbf{y} - \mathbf{h}(\mathbf{x}) \right)$

Evaluation obs O-B, O-A for obs network 2

60

50 40 30

20

10

60

50

20

10

O-B for Surf eval assim mean=-1.81118, stddev=22.94635, Nobs=184



Standard deviation of O-A for evaluation obs



Av lev height (kr

Diagonal B-matrix:

- Easy to represent
- Unrealistic, but potentially well conditioned numerically
- Lengthscales (ℓ) are the same as the grid boxes
- Need to know only background error variances
- Non-diagonal B-matrix:
 - Complicated/potentially expensive to represent
 - Need to know:
 - Error variances
 - Horiz. and vert. error corrs. in x^{ic}
 - Horiz. and temp. error corrs. in **x**^{sf}
 - Need to represent a big matrix:
 - Full B: [(64×32×60) + $(64 \times 32 \times 16)$]² $\approx 24 \times 10^{9}$ elements
 - Horiz. corrs. only (for SF): [64×32]² $\approx 4.2 \times 10^6$ elements
 - We test an efficient implicit representation of B via an efficient spectral decomposition method
 - Spectral representation: only 33 elements for horiz!

Experiments

- 665 surface obs.
- ~75% assimilated
- ~25% unassimilated (for evaluation)
- Above chosen randomly three times (obs networks 1, 2, 3)
- Different B-matrix configs. (see key)

