

The GMAO 4DVAR and its Adjoint Tools

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Abstract

The fifth generation of the Goddard Earth Observing System (GEOS-5) Data Assimilation System (DAS) is a 3d-var system that uses the Grid-point Statistical Interpolation (GSI) system developed in collaboration with NCEP, and a general circulation model developed at Goddard, that includes the finite-volume hydrodynamics of GEOS-4 wrapped in the Earth System Modeling Framework and physical packages tuned to provide a reliable hydrological cycle for the integration of the Modern Era Retrospective-analysis for Research and Applications (MERRA). This MERRA system is essentially complete and the next generation GEOS is under intense development. A prototype next generation system is now complete and has been producing preliminary results. This prototype system replaces the GSI-based Incremental Analysis Update procedure with a GSI-based 4d-var which uses the adjoint of the finite-volume hydrodynamics of GEOS-4 together with a vertical diffusing scheme for simplified physics. As part of this development we have kept the GEOS-5 IAU procedure as an option and have added the capability to experiment with a First Guess at the Appropriate Time (FGAT) procedure, thus allowing for at least three modes of running the data assimilation experiments.

The prototype system is a large extension of GEOS-5 as it also includes various adjoint-based tools, namely, a forecast sensitivity tool, a singular vector tool, and an observation impact tool, that combines the model sensitivity tool with a GSI-based adjoint tool. These features bring the global data assimilation effort at Goddard up to date with technologies used in data assimilation systems at major meteorological centers elsewhere.

Four-dimensional Variational Approach

The general cost function of the variational formulation

$$J(x) = \frac{1}{2} (x_0 - x^*)^T B^{-1} (x_0 - x^*) + J_0 + \frac{1}{2} \sum_{k=1}^K [(x_k - y_k)^T R_k^{-1} (x_k - y_k)] + \frac{1}{2} \sum_{k=1}^K [(x_k - x_{k-1})^T Q_k^{-1} (x_k - x_{k-1})]$$

where

- $x \equiv [x_0, x_1, \dots, x_K]^T$ is a 4d state vector;
- h_k and m_k are the nonlinear observation and dynamical model operators, respectively;
- B , Q_k , and R_k are the background, model, and observation error covariances, respectively;
- Strong constraint formulation: $Q_k \rightarrow \infty$;
- Weak constraint formulation: $Q_k \neq 0$ accounts for imperfections in the model an;
- J_0 represents a balance constraint.

Incremental Variational Formulation

The minimization of J is iteratively treated in the Gauss-Newton sense where an iterative procedure linearizes the cost function at each, so called, outer loop, turning the problem in to a quadratic minimization problem for the following function

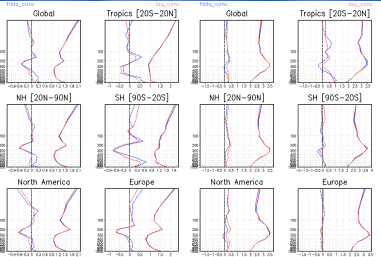
$$J_i \equiv J(\delta x_i) = \frac{1}{2} (\delta x_i - \delta x_i^*)^T B^{-1} (\delta x_i - \delta x_i^*) + \frac{1}{2} (H_i \delta x_i - d_i)^T R_i^{-1} (H_i \delta x_i - d_i)$$

- where $\delta x_i \equiv x - x_{i-1}$, $\delta x_i^* \equiv x^* - x_{i-1}$, and
- $\delta x_i \equiv x_i - x_{i-1}$ is the control variable;
- R is a 4d matrix combining the matrices R_0 and Q_k ;
- The inner loop minimization of J_i can be solved by
 - Conjugate gradient
 - Quasi-Newton (such as L-BFGS)
 - Lanczos
- Conditioning of the J_i minimization is determined by the Hessian $\nabla^2 J_i \equiv B^{-1} + H_i^T R_i^{-1} H_i$, which spectrum is such that a good preconditioner is essential, particularly in 4dvar.

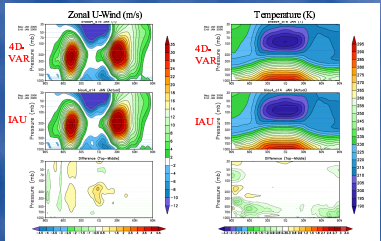
Experimental Setup IAU and 4DVAR to GEOS-5

- 20207 resolution
- Period: January 2008
- Companion based on North Means and Radial Statistics
- Both 3dvar and 4dvar use Lanczos-based Q2, S2000, B1000s

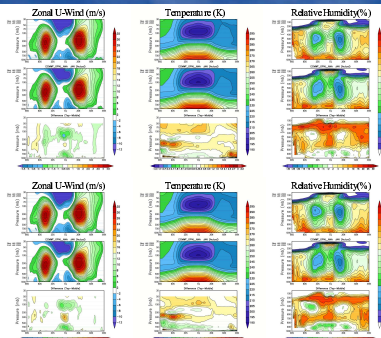
The first test of the newly implemented 4DVAR is to compare its results against the present GMAO IAU-based 3dvar assimilation system. Using the set up above, we ran an experiment for January 2008 and illustrate the comparison with IAU in the following few figures. The statistics of various observation-minus-background residuals (see model errors) to produce. The figure below shows zonal wind (0 bands on the left) and temperature residuals (0 bands on the right) for radiances. Introduction of 4DVAR affects mainly the biases in these residuals and biases. In the case of the winds, mid-tropospheric wind and temperature biases seem to improve slightly with 4DVAR, with a small degradation of the low-tropospheric temperatures.



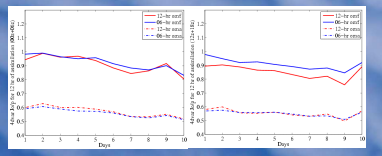
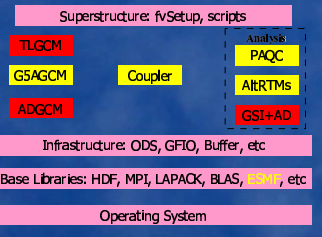
In the zonal-averaged monthly mean sense IAU and 4DVAR behave fairly similarly at first glance. Closer examination of the difference shows more explicitly the differences in bias between IAU and 4DVAR. This is illustrated in the figure below for zonal wind (left) and temperature.



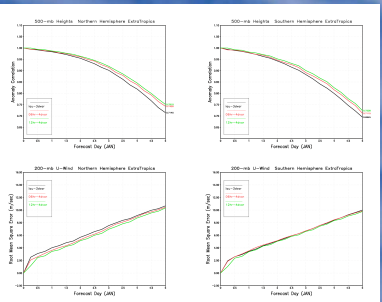
Though the observation-minus-background residual statistics is already very telling for when and where 4DVAR is an improvement over IAU, or is a small addition, comparison against other center biases can help further. Below, we compare the zonal-averaged monthly means of zonal wind, temperature and relative humidity with ECMWF operational fields for IAU (top) and 4DVAR (bottom).



First, we note that at this low 2-dg resolution the GEOS-5 IAU-3dvar system has a strong bias in the tropical zonal wind (not necessarily typical of the GEOS-5 full resolution system), atmospheric mid-troposphere wind bias is also present to a lesser extent. Similarly, IAU-3dvar shows somewhat negative and positive atmospheric bias in the temperature field. The comparison with ECMWF seems to corroborate the residual statistics results showing improved tropical tropospheric winds, indeed even the mid-troposphere winds seem to improve considerably with 4DVAR. The temperature bias also improves in the low-troposphere, mid-troposphere tropics and NH, with some deterioration over the mid-troposphere SH mid-troposphere tropics. The relative humidity fields are consistent with the temperature changes.



While extending the time-window of 4DVAR from 6 to 12-hr we look at the normalized observation cost function, indicative of the goodness of fit, before the minimization starts and at the end of the minimization. A body line series of plots is shown above for both the 6- and 12-hr 4DVAR. To make the proper comparison we also show the results for two consecutive 6-hr windows for the 6-hr 4DVAR. The results are an indication that in the average the assimilation convergence is similar for both cases. Indeed, the 12-hr 4DVAR seems to give slightly enhanced fits for most of the stations.



Finally, we show results of forecast skill scores for IAU and both 4DVAR setups using either a 6- or a 12-hr time window. We remind the reader that these scores are for a low resolution GEOS-5 DAS and do not reflect the actual scores when the full resolution DAS is used. The figures above confirm what the statistics and the comparison with ECMWF operational analysis indicate, that is, that the 6-hr 4DVAR is indeed an improvement over IAU-3dvar particularly in the NH subtropics. Furthermore, the figures also show that the 12-hr 4DVAR provides either neutral or improved scores over the 6-hr 4DVAR results. Though preliminary, these results are very encouraging.

Observation Sensitivity and Impact

Take \mathcal{F} to be the measure of an aspect of the forecast one wishes to examine. Since the forecast \hat{f} is a function of the background field x^* and the observations y , this measure is a convolution of operations:

$$\mathcal{F}(x^*, y) = F * M * G(x^*, y)$$

where M and G represent the forecast model and the data assimilation system, respectively.

The sensitivity of the forecast to observations is

$$\frac{\partial \mathcal{F}}{\partial y} = G^T M^T \frac{\partial F}{\partial x^*} = G^T \frac{\partial F}{\partial x^*}$$

which in general requires second order adjoint information (Le Dimet et al 2002).

The observation impact, defined as the change in the forecast aspect \mathcal{F} due to a set of observations can be approximated to first order by

$$I_1 \equiv \frac{\partial \mathcal{F}}{\partial y} \delta x >$$

where δx represents an analysis increment.

For a linear DAS, $\delta x = Kd$, here,

$$I_1 \equiv \frac{\partial \mathcal{F}}{\partial x} Kd > \equiv K^T \frac{\partial \mathcal{F}}{\partial x^*} d > \equiv \frac{\partial \mathcal{F}}{\partial x^*} d >$$

For a nonlinear DAS, the increment is a successive correction of the linear-type increment, that is,

$$\delta x_i = K_i d_i + (I - K_i H_i)(x^* - x_{i-1})$$

The final (total) increment is

$$\delta x = \sum_{j=1}^n K_j H_j d_j + \dots + K_{n+1} H_{n+1} K_n \frac{\partial \mathcal{F}}{\partial x^*} d_1 >$$

Therefore, for a nonlinear DAS, the first order impact is approximately

$$I_1 \equiv \sum_{j=1}^n \langle K_n H_n \dots K_{j+1} H_{j+1} K_j \frac{\partial \mathcal{F}}{\partial x^*} d_j >$$

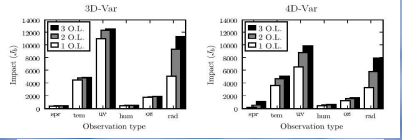
Higher order terms for linear DAS have been derived and discussed in Errico (2007) and Gelaro et al. (2007); Tremolet (2008) gives a comprehensive discussion and derivation for nonlinear DAS.

The Adjoint of a Variational Analysis System

Observation sensitivity and impact studies require the adjoint of the underlying data assimilation system. The model adjoint provides the model sensitivity as the input to the analysis adjoint for calculation of the observation sensitivities.

Concentrating on the analysis adjoint, there are at least three ways to obtain the adjoint of a variational analysis system:

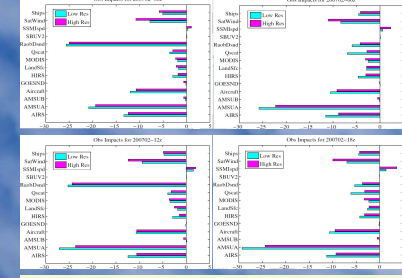
- Direct, line-by-line, adjoint (Zhu & Gelaro 2007)
- Operator manipulation:
 - Observation space (Baker & Daley 2000): $K^T \partial \mathcal{F} / \partial x \equiv \partial z$ $(H^T H^T + R) z = H^T \partial \mathcal{F} / \partial x$
 - In physical/spectral space (Tremolet 2008): $K^T \partial \mathcal{F} / \partial x \equiv R^{-1} H \delta x$ $(I^T + H^T R^{-1} H) \delta x = \partial \mathcal{F} / \partial x$
 - Approximating the Hessian (Cardinali et al. 2004): $K^T \partial^2 \mathcal{F} / \partial x^2 \approx R^{-1} H \partial^2 \mathcal{F} / \partial x$ $\tilde{A} \sim \sqrt{B^T + H^T R^{-1} H}$



The figure above shows the observations impact on Jb for 100 inner-loop iterations and three outer loop iterations. Results for 3dvar (left) and 4dvar (right) are shown for each of the control variables. 4dvar outer loop have larger impact than 3dvar outer loops; the overall impact of 4dvar is smaller than that of 3dvar - 4dvar makes more consistent use of observations through the time window - Jb is smaller in 4dvar even though a larger number of observations end up being used (not shown).

Observation Impacts on 24hr forecast for resolutions of GEOS-5 DAS

- Two experiments at 202072
- and a third at 0.50, 0.25, 0.125 resolution
- Period: February 2007
- Each experiment verified against NCEP analysis



The figure above shows observation impacts on the 24hr forecasts from GEOS-5 DAS (0.50x0.25x0.125) over the period of February 2007. Results for a 6-hour window (left bar) and the full 24-hour window DAS are displayed. The impacts are presented into each of the four synoptic analysis times and grouped as some of the most revealing synoptic types currently present in GEOS-5. Typical statistics of multiple observations from these plots is that the observing systems of most significance to the 24hr forecast are the radiosonde and the AMSU radiances; the impact of AMSU radiances and satellite winds are also very significant. Moreover, one sees that AMSU radiances at times when the radiosonde observing network is less active (a 02z and 18z). Also, we see that the relative impacts of the various observing systems does not change significantly when resolution changes making it feasible to use lower resolution experiments to quickly assess overall impact of various new data types. Note: positive impacts on the 24hr forecast are negative numbers; negative impacts are positive numbers.

Summary and Conclusions

- NASA GEOS DAS has entered the 4D-world (again!)
- Various adjoint tools are now available in GEOS-DAS, capable of performing studies in forecast sensitivities, singular vectors, analysis sensitivity and observations impact
- First exercise including some of these tools will be the Observations Impact Inter-comparison Study (ECMWF, Env. Canada, Meteo France, NASA, and NRL)
- Hooks for weak constraint are in place in GSI and soon will be in place in the GEOS-5 GCM Work is on way to update the GCM TLM/ADM with cube-sphere core
- Soon we will be able to compare 4DVAR with NCEP's approximate 4D-scheme; First Order Time-Interpolation to Observations (FOTD)

Background photo from Earth Observation Station by Keith Miller, Apperian with the moon from Gaspard Cardou, Thierry Del...

The implementations done thus far benefited greatly from the incredible infrastructure of GSI.

International TOVS Study Conference, 16th, ITSC-16, Angra dos Reis, Brazil, 7-13 May 2008.
Madison, WI, University of Wisconsin-Madison, Space Science and Engineering Center,
Cooperative Institute for Meteorological Satellite Studies, 2008.