



# Regeneration of CO<sub>2</sub> Satellite Column Data tailored to an Atmospheric inversion Scheme

A. Webb<sup>1,2</sup>, H. Boesch<sup>1,2</sup>, F. Chevallier<sup>3</sup>, C. O'Dell<sup>4</sup>

Contact: alex.webb@leicester.ac.uk

<sup>1</sup>National Centre for Earth Observation, University of Leicester, Leicester, UK.

<sup>2</sup>Earth Observation Science, Department of Physics and Astronomy, University of Leicester, Leicester, UK.

<sup>3</sup>Laboratoire des Sciences du Climat et de l'Environnement, CEA-CNRS-UVSQ, L'Orme des Merisiers, Bat 701, 91191 Gif-sur-Yvette, France. <sup>4</sup>Colorado State University, Fort Collins, CO, USA.

#### Introduction

• Current atmospheric inversion schemes assume rather realistic prior information when projected in concentration space, while satellite retrieval schemes try to maximize the measurement contribution in the retrievals by giving a very weak weight to prior information.

• Such inconsistent statistical hypotheses between XCO<sub>2</sub> retrieval and atmospheric inversion schemes may be a significant cause of error in atmospheric inversions assimilating satellite data.

• By using sounding-specific covariances from model covariances representing the prior flux uncertainties of the MACC model projected in space and time, we can derive satellite CO<sub>2</sub> columns tailored towards and consistent with the specific assumptions of the MACC atmospheric inversion scheme.

• Using global level-2 CO<sub>2</sub> data from NASA's OCO-2 satellite we reconstruct CO<sub>2</sub> columns with prior information from the MACC model using linear approximations thus avoiding the need for time-consuming re-retrieval of the satellite observations.

## OCO-2 CO<sub>2</sub>

• NASA's OCO-2 satellite (Orbiting Carbon Observatory 2) was launched on July 2<sup>nd</sup> 2014 into a near-polar orbit, joining the A-Train formation of satellites.

#### MACC CO<sub>2</sub> Correlation Matrices

• The instrument incorporates three high-resolution spectrometers that make coincident measurements of reflected sunlight in the near-infrared CO<sub>2</sub> near 1.61 and 2.06  $\mu$ m, and in the O<sub>2</sub> A-Band at 0.76  $\mu$ m.

• CO<sub>2</sub> vertical profiles are retrieved using an optimal estimation scheme which combines OCO-2 radiance measurements with a priori information, a radiative transfer model and error covariance matrices associated with the measurements and a priori values.



Figure 1: OCO-2 L2 Diagnostic 8r XCO<sub>2</sub>, 2016-06, 2°x2° res.

#### Method

Retrieval schemes like ACOS usually try to follow Bayes' theorem in a Gaussian framework in order to find the most likely values of a series of variables gathered in a state vector **x** given:

- A prior value of **x**: **x**<sup>b</sup> and its error covariance matrix: **B**
- Satellite measurements **y** and their error covariance matrix **R**
- An operator H which links the two types of variables

If *H* is linear, the solution **x**<sup>a</sup> to the inverse problem and its error covariance matrix **A** are given by:

$x^a = x^b + K(y - Hx^b)$ ,	(1)
$A = (H^T R^{-1} H + B^{-1})^{-1}$ ,	(2)
A = (I - KH)B,	(3)

where **H** is the Jacobian matrix of *H*, **I** is the appropriate identity matrix and **K** is defined as:  $K = (H^T R^{-1} H + B^{-1})^{-1} H^T R^{-1}.$  (4)

We can rewrite (1) and use (2), (3) and (4) to show that,  $A^{-1}x^a - B^{-1}x^b = H^T R^{-1}y,$ 



• We use correlation matrices from the MACC CO<sub>2</sub> inversion model v14.2 which covers 1970-2014 at 3.75°x1.9° and 3-hourly resolution.

• The MACC inversion also relies on optimal estimation, using a priori, a transport model and error covariance matrices.

• The a priori error covariance matrices utilised by the model are constructed using a climatology as a function of month, latitude and surface type.

Figure 2 (left): Example MACC a priori error correlation matrices for  $CO_2$  for January (left), June (right), land (top), ocean (bottom) at 1° latitude. The top right represents the top of atmosphere, the bottom left represents the surface.

### **Results and Future Work**

Figure 4 (right) shows the original a priori correlation matrix from OCO-2 and the corresponding a posteriori correlation matrix for an example case alongside the MACC a priori error correlation matrix (interpolated onto the OCO-2 pressure profile) and the newly calculated a posteriori error correlation.





Figures 5, 6, 7 (left top to bottom) show: The original L2 OCO-2 data (June 2016), the recalculated OCO-2 and the difference (recalculated minus original).

0.4

0.6

0.8

NERC

SCIENCE OF THE ENVIRONMENT

Equation 5 implies that for given measurements  $\{y, R\}$  and for a given observation operator H, two retrieval schemes (noted 1 and 2) that differ by their a priori information  $\{x_1^b, B_1\}$  and  $\{x_2^b, B_2\}$  respect the equality:

(5)

$$A_1^{-1}x_1^a - B_1^{-1}x_1^b = A_2^{-1}x_2^a - B_2^{-1}x_2^b, \tag{6}$$

So that:

and:

$$x_{2}^{a} = A_{2} \left( A_{1}^{-1} x_{1}^{a} - B_{1}^{-1} x_{1}^{b} + B_{2}^{-1} x_{2}^{b} \right), \qquad (7)$$
$$A_{2} = \left( A_{1}^{-1} - B_{1}^{-1} + B_{2}^{-1} \right)^{-1}. \qquad (8)$$

Equations 7 and 8 are entirely in the space of the state vector, allowing us to change the a priori error covariance matrix and obtain a new solution to the inverse problem without requiring the observations, their error covariance matrix or the observation operator.

These two equations have the clear advantage (over equations 1 and 2) of limiting the size of matrices in solving the retrieval.

Figure 8 (below) shows time-series of the
daily averaged XCO<sub>2</sub>: shaded area showing
the standard deviation (std) of the data. The
recalculated dataset is in blue and the
original OCO-2 data in grey.



Acknowledgements

The OCO-2 data were produced by the OCO-2 project at the Jet Propulsion Laboratory, California Institute of Technology, and obtained from the OCO-2 data archive maintained at the NASA Goddard Earth Science Data and Information Services Center.

