

MIPAS-ENVISAT AND IASI-METOP DATA FUSION USING THE MEASUREMENT SPACE SOLUTION METHOD

Samuele Del Bianco, Ugo Cortesi, Simone Ceccherini, Piera Raspollini

Istituto di Fisica Applicata "Nello Carrara" del Consiglio Nazionale delle Ricerche (IFAC-CNR)

Introduction

Techniques of data fusion are presently being considered with increasing interest for application to atmospheric observations from space because of their capability to optimally exploit the complementary information provided by concurrent instruments operating aboard on-going and future satellite missions.

The task of combining individual measurements of the same target, when carried out at level of the retrieved atmospheric state vectors generally requires the interpolation of products represented on different retrieval grids and the use of a priori information that can lead, respectively, to a loss of information and to an introduction of biases in the fused data.

In order to avoid this kind of degradation in the guality of the final products, a new approach to the problem of data fusion has recently been proposed relying on the Measurement Space Solution (MSS) method (Ceccherini et al., 2009).

Here we present a first application of the MSS method to the problem of combining measurements of ozone total and partial columns obtained by two FTS instruments: the MIPAS limb sounder onboard the ENVISAT satellite and the IASI nadir-viewing spectrometer onboard the METOP-A platform.

The results are based on simulated observations and show the evidence of improved retrieval quality, in terms of retrieval errors and averaging kernels, when comparing the outcome of data fusion with those of the inversion process applied to spectra from either of the two instruments.

the proposed strategy for data fusion to MIPAS and IASI real observations is

The Measurement Space Solution method

The MSS method makes possible optimal exploitation and representation of the information retrieved from atmospheric soundings, by separating the information coming from the measurements and that obtained from external constraints.

The atmospheric state vector \mathbf{x} is decomposed in its components \mathbf{x}_{a} and \mathbf{x}_{b} belonging respectively to the measurement space (the space generated by the rows of the Jacobian matrix of the forward model) and to the null space (the orthogonal complement space):

 $\mathbf{x} = \mathbf{x}_a + \mathbf{x}_b$

where \mathbf{x}_{a} and \mathbf{x}_{b} can be determined In terms of the matrices V and W, whose columns are orthonormal bases of the measurement space and of the null space respectively, and of the projections a and b of x on these orthonormal bases:

x _a = V a	a = V ^T x
x _b = W b	b = W ^T x

As detailed in the paper by (Ceccherini et al., 2009) and briefly recalled by the same authors in their contribution to this poster session, the MSS $\mathbf{x}_{\mathbf{a}}$ can be determined using a singular value decomposition, to calculate the V matrix and to obtain an estimate a of the projection a.

Contributions of the measurement space and null space components of the profile to column

The MSS method can be used to express the contribution to total or partial columns of the components of the profile in the measurement space and in the null space The ozone column between altitude levels z_1 and z_2 can be written in discrete form as the scalar product:

 $c = v^T x$

where we have considered a segmentation of the atmosphere between z_1 and z_2 in N homogeneous layers of equal thickness Δz_i (with pressure p_i , temperature T_i and VMR x_i constant within each single layer) and where the components of **v** represent the partial columns of air within each layer. $v_i = \frac{T_i}{L_i} \frac{T_i}{T_i}$ (i = 1, ..., N) of air within each layer. $k T_i$

By using the decomposition of x in its measurement space and null space components, we obtain :

where c_a and c_b are the contributions to the column from the components of the profile in the measurement space and in the null space, respectively:

$$c_a = \mathbf{v}^T \mathbf{x}_a = \mathbf{v}_a^T \mathbf{x} \qquad (\mathbf{v}_a = \mathbf{V} \mathbf{V}^T \mathbf{v})$$

$$\mathbf{v}_{\mathsf{B}} = \mathbf{v}^{\mathsf{T}} \, \mathbf{x}_{\mathsf{B}} = \mathbf{v}_{\mathsf{b}}^{\mathsf{T}} \, \mathbf{x} \qquad (\mathbf{v}_{\mathsf{b}} = \mathbf{W} \, \mathbf{W}^{\mathsf{T}} \, \mathbf{v})$$

with v_a and v_b components of v in the measurement space and in the null space, respectively.

An estimate of the some on the total error ϵ is given by: $\begin{aligned} & |\hat{c}_a = v^T V \, \hat{a} \\ & |\hat{c}_a = v^T V \, \hat{a} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ & |\hat{c}_a = \sqrt{v^T V \, S_a v^T v} \\ &$

Application of the MSS method to data fusion

In the case that two or more instruments sound the same portion of atmosphere the information contained in the observations of all the instruments can be exploited to obtain a more precise estimation of the ozone column with respect to when a single instrument performs the measurement.

Here we follow the approach to data fusion proposed by (Ceccherini et al., 2009), where starting from the MSSs of the ozone profile of the individual measurements a new MSS is calculated which lies in the union space of the original measurement spaces. Once that this new MSS has been calculated (corresponding to a new matrix V and a new vector â) the above described procedure can be applied to determine a new estimation of the ozone column that includes the information coming from all the considered measurements

This approach has been applied to combine the ozone measurements from MIPAS-ENVISAT

Retrieval of O₃ partial and total columns from simulated observations

In Fig. 1, we have reported the percentage difference between the retrieved and the true values of tropospheric, stratospheric and total ozone columns as a function of the number of singular values used for the determination of the measurement space. In the upper panels results of the retrieval from IASI measurements (a), from MIPAS measurements (b) and from IASI/MIPAS data fusion (c) are displayed.

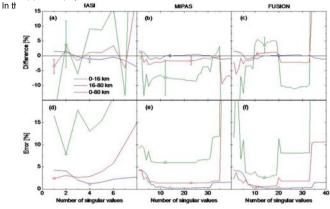


FIGURE 1

The number of singular values corresponding to the minimum of the total error is highlighted in the upper panels with symbols associated to the specific partial or total columns (triangle for tropospheric, diamond for stratospheric and square for total column).

In Table 1 percentage errors $\epsilon,~\epsilon_a$ and $\epsilon_b,$ calculated for the number p of singular values that corresponds

	Tropospheric column [0 - 16] km				Stratospheric column [16 - 80] km				Total column [0 - 80] km			
	р	ε.[%]	£₅[%]	ɛ[%]	p	£,[%]	8 _b [%]	ɛ[%]	p	e ,[%]	Eb[%]	ɛ[%]
IASI	2	2.52	7.53	7.94	4	0.77	0.85	1.15	1	0.09	2.34	2.34
MIPAS	12	0.80	5.94	5.99	14	0.16	0.17	0.23	23	0.19	1.33	1.35
Fusion	14	1.20	2.28	2.58	20	0.17	0.15	0.22	11	0.25	0.41	0.48

TABLE 1

As a result of data fusion a significant reduction in the total error is obtained particularly in the

Sensitivity of the retrieved columns to the true ozone distribution

In order to characterize the sensitivity of the estimated column c to the true ozone distribution and to evaluate its improvement with data fusion, we have considered the averaging kernel (AK) in the partial column space.

This is defined as the vector A, whose ith element is the derivative of the estimated column with respect to the partial column $c_i = v_i x_i$ of the ith layer, given

$$A = \frac{V_{ai}}{V_{ai}}$$

$$A_i = \frac{u}{V_i}$$

The expression above can be derived by writing the derivative of the estimated column with respect to the true ozone VMR profile as:

$$\frac{\partial \hat{\mathbf{c}}}{\partial \mathbf{x}_{i}} = \mathbf{v}_{ai}$$

or, equivalently, as

by:

$$\frac{\partial \hat{\mathbf{c}}}{\partial \mathbf{x}_{i}} = \sum_{j=1}^{N} \frac{\partial \hat{\mathbf{c}}}{\partial c_{j}} \frac{\partial c_{j}}{\partial \mathbf{x}_{i}} = \frac{\partial \hat{\mathbf{c}}}{\partial \mathbf{c}_{i}} \mathbf{v}_{i} = \mathbf{A}_{i} \mathbf{v}_{i}$$

In Fig. 2, we report the averaging kernels A for the retrieval of tropospheric, stratospheric and total ozone columns, when using:

(a) IASI measurement only

(b) MIPAS measurement only

(c) IASI and MIPAS data fusion

The AKs of the individual measurements deviate substantially from the ideal behavior (AK equal to 1 within the altitude range of the partial column under consideration and zero outside). The data fusion improves the sensitivity below 6 km w.r.t. MIPAS measurements and above 6 km w.r.t. IASI measurements.

In Fig. 3, we illustrate the dependence of the AKs on the number of considered singular values, showing the AKs of the total column (0-80 km) retrieved for the data fusion of MIPAS and IASI measurements for several values of the number p of considered singular values. For increasing values of p the AK approaches the behavior expected in the ideal case (at expenses of an increase of the total error)

The choice of the optimal p value is driven by the trade-off between the shape of the AK and the

References

S. Ceccherini, P. Raspollini, B. Carli, Optimal use of the information provided by indirect measurements of atmospheric vertical profiles, Opt. Express, Vol. 17, n. 7, 4944 - 4958, 2009.

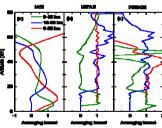


FIGURE 2

E 0 ø Ø. • 1 FIGURE 3