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Probing ETEX-II data set with inverse modelling

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Abstract

We give here an account on the results of source inversion of the ETEX-II experiment. Inversion has been performed with the maximum entropy method on the basis of non-zero measurements and in conjunction with a transport model POLAIR3D. The discrepancy scaling factor between the true and the reconstructed mass has been estimated to be equal to 7. The results contrast with the method’s performance on the ETEX-I source. In the latter case its mass has been reconstructed with an accuracy exceeding 80%. The large value of the discrepancy factor for ETEX-II could be ascribed to modelling difficulties, possibly linked not to the transport model itself but rather to the quality of the measurements.

1 Introduction

The European Tracer EXperiment (ETEX) was designed to evaluate forecast performance of dispersion models as well as readiness and communication capacity of European institutions in case of an accidental release of harmful substances to the atmosphere. The experiment consisted of two controlled releases of harmless and non-depositing inert tracers (perfluorocarbon compounds). Since the release term was controlled and thereby well-known, and since the deposition and chemistry processes were non-existing, the design of the experiment made it an excellent test on models of atmospheric transport and diffusion from point sources at continental scale.

The location of the two releases was in Monterfil, Brittany, France. The first release (ETEX-I) took place on 23 October 1994 and the second release (ETEX-II) some three weeks later, on 14 November 1994. Both experiments were carried out under comparable meteorological conditions with a low pressure system situated north of Scotland and westerly wind over the release site and over Western and Central Europe. In this paper we focus on the second ETEX release which took place between 15:00 UTC on 14 November and 02:45 UTC on 15 November 1994. 490 kg of PMCP (perfluor-
romethylcyclopentane) were released with a constant rate. In the first stage after the release a narrow cloud was advected rapidly eastwards (see the modelled plume in Fig. 1). Advection was driven by strong winds preceding a frontal passage which was linked with the low pressure area moving from the North Atlantic towards Scandinavia and Russia. The wind during ETEX-II was, however, much stronger compared to the ETEX-I episode, and the low pressure system over Western Europe was accompanied by heavy rain. The experimental campaign coincided with a modelling exercise in which 28 models took part. Modelling turned out to be more difficult than in the case of the first release (Brandt, 1998).

The first ETEX experiment has been treated in many papers in which model results have been compared to observations, see (e.g., Brandt et al., 1998). The ultimate evaluation of the outcome of model intercomparison was presented in the final reports in 1998 (Mosca et al., 1998; Graziani et al., 1998a) where results from 49 models were included. In (Mosca et al., 1998) it was concluded that: “almost all the models show a satisfactory agreement with the measured values”.

Even though the meteorological conditions were quite similar, and the same measurement sites and procedures were used in both experiments, there was a significant difference in models’ performance between ETEX-I and ETEX-II. In the latter case an overall bias of a model validated against the measurements is suggested to be of a factor of 10 (Brandt, 1998). The experiment’s website (http://rem.jrc.cec.eu.int/etex/) reports spatial and temporal analyses for a very limited number of time instants and monitoring stations. Figures of merit and time of arrival for some subsets of models are the only statistical indices disclosed by the JRC. In particular, a gap in the quality of model results for the two releases is underlined there. For the figure of merit in space (FMS) at 24 hours after the release, half of the models exceeded 20% but only one of them reached the value of 40%. In the case of the first ETEX release, an FMS above 40% was attained by 4 models out of 28, and there were still models exceeding this value at 36 h and even 60 h after the release.

According to the detailed conclusions from the ETEX symposium in 1997 in Vienna,
the following was stated about the ETEX-II release: “The second release showed larger discrepancies between observations and model results where all models significantly overpredicted surface concentrations after 12 h from the start of release. No clear explanation has yet been given for this result” (Nodop, 1997). The eventual evaluation of model intercomparison within ETEX-II was treated in the final report in 1998 (Graziani et al., 1998b) where results from the 28 models were presented. Some of the conclusions in this report stated: “Compared to the first release, much less tracer was found”. It was also reported that: “Not only fewer non-zero values were observed, but also the concentration levels were much lower”. Furthermore, according to the conclusions: “the cloud is apparently broken up into several parts moving at different speed in different directions. At many sites, the tracer was detected intermittently, rather than in a continuous sequence, as it might be expected”. In the end, the report settles that: “The generally poor performance of the long-range dispersion models employed in the real-time simulation of the release are to put in relation with the complexity of the weather situation after the end of the second ETEX experiment”.

Bad models’ performance was considered rather strange among the modelling community because all the models performed satisfactorily with respect to ETEX-I. Since all the models showed a large overestimation with respect to ETEX-II, the reason for this overprediction could be explained by one of the following two hypotheses: 1) there is a systematic error in all the model results e.g. due to the complexity of the weather situation which could result in a strong vertical mixing, or 2) there is a systematic error in the measurements which results in a significant negative bias (i.e. all the measurements are too small). The first hypothesis was concluded as a fact in (Graziani et al., 1998b). In this study, an inverse modelling procedure that was evaluated against ETEX-I with very good results (Bocquet, 2007a) suggests that the second hypothesis is more likely to be true.
2 Inverse modelling procedure

We tackle in this work the reconstruction of the source of ETEX-II in order to check to what extent the retrieved temporal profile and the emitted mass agree with the true release. Source retrieval problem is addressed with the principle of maximum entropy on the mean. The method has been presented in (Bocquet, 2005a,c), and successfully applied to inversion of the source of ETEX-I (Bocquet, 2007a) as well as the temporal and vertical profile of the source of the Chernobyl accident (Davoine and Bocquet, 2007).

2.1 Measurements

There are 2248 valid measurements collected in the experiment. In contrast, however, with 969 non-zero values gathered in ETEX-I, only 476 measurements (12% of all the measurements) resulting from the second release are non-null. Solely the latter subset has been used to perform inversion. The reason is our previous experience with the inversion method which made it clear that adding additional zero measurements is inefficient (Bocquet, 2007a).

2.2 Inversion method

The retrieval of a source from measurements sums up to an inversion of a source - measurement relationship. For a discrete formulation of the problem the relationship is of the form:

$$\mu = H\sigma + \epsilon,$$

(1)

where $\sigma \in \mathbb{R}^N$ is a source vector, and $\mu \in \mathbb{R}^d$ stands for the measurements. The matrix $H$ (known as the Jacobian matrix) embodies modelling of dispersion phenomenon as well as sampling of measurements. Therefore, $\epsilon \in \mathbb{R}^d$ represents a combined model-representativity-measurement error. Due to a large number of unknowns $N$ that we
want to invert and a comparatively small number of measurements $d$, $H$ has been computed in an adjoint mode.

### 2.2.1 Duality test

The modelling of the transport phenomenon for the needs of the inversion procedure is ensured by our Eulerian chemistry transport model POLAIR3D. It is equipped with a module devoted to radionuclides which has recently been validated on ETEX-I, Chernobyl and Algeciras releases (Quélo et al., 2007).

There are two ways of obtaining an adjoint associated with a discrete direct problem. One can discretise an adjoint of an original continuous model or adjoin an already discretised one. We have chosen here the first option which has the advantage to be simple. It boils down to time reversal of meteorological fields in a direct model. The latter possibility could have also been chosen since POLAIR3D has an extensively tested and employed adjoint module. In that case an agreement between the synthetic measurements obtained via the adjoint model, $c_i^{\text{adjoint}}$, Eq. (1) and the corresponding ones delivered by the direct simulation, $c_i^{\text{direct}}$ would have been precise. The precision being obviously up to computer’s accuracy, provided the dispersion model is linear. Our choice of the adjoint implies a poorer match between the values of synthetic measurements computed in these two different ways. The discrepancy is shown in Fig. 2 and contrasts with a straight line that would have been obtained if we had used the adjoint of a discrete model. To estimate quantitatively the discrepancies, we have computed the average fractional bias between both types of forecast concentration, $c_i^{\text{direct}}$ and $c_i^{\text{adjoint}}$:

$$
2 \frac{D}{\sum_{i=1}^{D} \left| \frac{c_i^{\text{direct}} - c_i^{\text{adjoint}}}{c_i^{\text{direct}} + c_i^{\text{adjoint}}} \right|.
$$

(2)

Whenever both of them are below 0.01 ng m$^{-3}$, it has been assumed that they match.
perfectly (negligible concentration is assessed in both cases). The mismatch indicator computed with respect to all the measurements, \( D = 2248 \), is greater then 20% for 366 measurements (that is to say for 16.3% of the concentrations), whereas 112 measurements (5%) have a mismatch exceeding 50%.

We deal in this paper, however, with an inversion of real measurements which are subject to measurement and modelling error, Eq. (1). In comparison to them, the error linked to the way the adjoint solutions have been computed can be regarded as negligible.

### 2.2.2 Cost function

The inversion method allows to take into account some prior knowledge which is appropriate to an accidental source. This capacity is of crucial importance due to scarcity of information on an accidental release and hence the necessity of making use of any available pieces of it. Prior information to be exploited refers in particular to positivity and boundedness of the source. Mathematical form of such information is encoded in a cost function arising from, for instance, Bernoulli law. However, any other positive prior law, such as an exponential law, would have served a decent purpose. The minimum of a cost function allows to shift from prior \( (\nu(\sigma) \text{ for the source and } \zeta(\varepsilon) \text{ for the errors}) \) to posterior probability density function \( (p(\sigma) \text{ and } q(\varepsilon), \text{ respectively}) \), and consequently defines the inverted source. Usually such a cost function is defined in the space of the discretised source \( \sigma \in \mathbb{R}^N \) (and errors \( \varepsilon \in \mathbb{R}^d \)):

\[
K_{\sigma,\varepsilon} = K_\sigma + K_\varepsilon = \sum_\sigma p(\sigma) \ln \frac{p(\sigma)}{\nu(\sigma)} + \sum_\varepsilon q(\varepsilon) \ln \frac{q(\varepsilon)}{\zeta(\varepsilon)},
\]  

and minimised under the constraint, Eq. (1). Because the number of observations \( (d = 476) \) is much lower than the number of unknown variables \( (N \approx 20 \times 10^3) \), it is computationally efficient to trade a cost function in the state space of the sources to a cost function in a space isomorphic to the observation space. Mathematics authorises the trade on the basis of the convexity of the cost functions which is guaranteed by the
maximum entropy on the mean approach. Ultimately, one deals with the following cost function (Bocquet, 2005b; Krysta and Bocquet, 2007):

$$\hat{L}(\beta) = \sum_{k=1}^{N} \ln \{1 - \gamma_k + \gamma_k \exp \left( m_k [\beta^\dagger H]_k \right) \} + \frac{1}{2} \beta^\dagger R \beta - \beta^\dagger \mu$$

(4)

where $\beta \in \mathbb{R}^d$ is the vector conjugated to the measurement vector $\mu$. $R$ is the observation error covariance matrix which parametrises the Gaussian prior law on the errors, $\zeta(\varepsilon)$. It is chosen diagonal, of the simplest form $R=\chi I$, with $I$ the identity matrix in observation space. The estimated source and errors are then given, in terms of the minimum $\overline{\beta}$ of $\hat{L}$, by the estimators:

$$\overline{\sigma}_k = m_k \frac{\gamma_k \exp \left( m_k \left[ \overline{\beta}^\dagger H \right]_k \right)}{1 - \gamma_k + \gamma_k \exp \left( m_k \left[ \overline{\beta}^\dagger H \right]_k \right)}$$

$$\overline{\varepsilon} = R \overline{\beta}.$$  

(5)

Note that $m_k$ is a mass scale that is taken constant on all grid-cells, $m_k=m$. $\gamma_k$ is the probability that a release occurs in cell $k$, which is also taken constant, $\gamma_k=\gamma$, and set to a very low value.

2.2.3 L-curve estimates

The prior probability density functions, $\nu(\sigma)$ and $\zeta(\varepsilon)$, are defined with the help of some parameters whose values are set arbitrarily according to experience, rather than to some objective criteria. In particular, for Bernoulli law, the emitted mass $m$ is an example of such a parameter. Similarly, Gaussian distribution describing errors is parametrised with an arbitrary value of the standard deviation $\sqrt{\chi}$. Therefore, in the first place, we establish the optimal values of those parameters. Due to nonlinearity of the dependence of the cost function on $m$, the analysis has been undertaken with the L-
curve technique. The technique was introduced in (Hansen, 1992), and its application to source inversion is detailed in (Davoine and Bocquet, 2007).

Firstly, a standard deviation uniform for all the measurements, and hence giving more weight to large values, has been chosen equal to $\sqrt{\chi}=0.3$ ng m$^{-3}$, according to (Bocquet, 2007a). While varying $m$ within a large interval of its values, we have checked for the one corresponding to the L-curve corner. It turned out to be $m=10 \times M/\lambda^2$, where $M$ is the true quantity of the released PMCP during a one-hour interval and $\lambda$ is a parameter that sets the space scale (arbitrarily equal to 4 and 8 for the resolution $2.25^\circ \times 2.25^\circ$ and $4.5^\circ \times 4.5^\circ$, respectively).

Then, having set $m=10$ (dropping the conventional unit $M/\lambda^2$), we have applied the procedure again, this time for a varying standard deviation. The point of inflection has been obtained for $\sqrt{\chi}=0.25$ ng m$^{-3}$, Fig. 3. Next, an L-curve experiment with a varying $m$ has been repeated, this time for the fixed $\sqrt{\chi}=0.25$ ng m$^{-3}$. It let us identify $m=8$ as the point of inflection, Fig. 4. At this point the iterative procedure has already converged to the optimal values of $m$ and $\chi$.

3 Inversion results for ETEX-II

We present in this section the inversion results obtained for the ETEX-II experiment.

3.1 Plain results

Fig 6 shows the inverted profile in Monterfil (in the grid-cell containing Monterfil) for the spatial resolution of $2.25^\circ \times 2.25^\circ$ and temporal resolution of 1 h, while Fig. 5 illustrates spatial distribution of the inverted source which has been integrated in time. The reconstructed mass in Monterfil is equal to 51 kg and the total reconstructed mass is 73 kg.

If we coarsen the spatial resolution of the domain of inversion, $4.5^\circ \times 4.5^\circ$, then the inverted mass in Monterfil is equal to 66 kg and the total reconstructed mass sums up
to 87 kg. On the other hand, inversion on the 1.125°×1.125° grid fails but, as shown in (Bocquet, 2005b), such a failure is bound to occur for a fine grid.

3.1.1 Marginal gain of information provided with by each observation

The assimilation of the data through the source inversion process results in a net gain of information accumulated by the optimality system. What the observations convey to the optimality system is measured by a relative entropy number denoted $K_\sigma$, Eq. (3). In a general, non-Gaussian case, a single measurement contribution to the gain $K_\sigma$ depends on the solution and hence on the actual values of the other measurements. That is why one investigates the marginal contribution $\partial_{\mu_i}K_\sigma$ of an observation $\mu_i$, for all the observations. In (Bocquet, 2007b), it was shown how to compute such a marginal contribution. These contributions have also been computed for ETEX-II and are plotted in Fig. 7.

Moreover, the Lagrange multipliers $\beta$ have been pictured. They represent the marginal total gain of information (which can either serve to identify the source or be lost in the deblurring of the errors in the data): $\partial_{\mu_i}K_{\sigma,\varepsilon}$. They are also a direct measure of how sensitive the optimality system is to each measurement seen as a constraint. Finally, because the prior hypothesis on the errors is Gaussian, they are also directly proportional to the diagnosed errors ($\varepsilon=\beta\beta^T$).

From Fig. 7, one learns that the diagnosed errors depend strongly on most measurements in France and Benelux: a change in a measurement results in an increase of the errors, rather than a new piece of information on the source. This contrasts with the analogue result for ETEX-I (Bocquet, 2007b). Nor is it the case for ETEX-II data coming from Central and Eastern Europe. The information that is gained on the source when perturbing the latter measurements is rather strong (again as compared to ETEX-I). This can be explained by the fact that the plume moved eastwards rapidly so that these observations are more informative than for ETEX-I. But also by the fact that the information that they bring about on the source is more trusted than the infor-
mation got from the Western measurements (at least relatively to the similar ETEX-I analysis). So that, comparing to ETEX-I, the optimality system trusts more the Eastern observations, which were probably found more consistent than the Western ones.

3.2 Comparison to ETEX-I

We report here inversion results as have been obtained for the ETEX-I experiment (Bocquet, 2007a), but improved by parameters’ estimation. In the first ETEX release 340 kg of PMCH were emitted to the atmosphere. Before source inversion has been addressed, a similar L-curve analysis has been performed. Consequently, an entire interval of the values of $m$ and $\chi$ ensuring robust inversions has been identified. The borders of the domain of those values are given by $[m=5; \chi=0.30 \text{ ng m}^{-3}] \times [m=12; \chi=0.45 \text{ ng m}^{-3}]$, where $m$ is in units of the truly released quantity of PMCH per hour. The reconstructed mass in Monterfil is 244 kg and the total reconstructed mass is equal to 344 kg for the lower boundaries and, respectively, 224 kg and 336 kg for the upper boundaries of the domain.

A noteworthy feature of the ETEX-I inversion is also the fact that, contrary to the ETEX-II results, the method lets invert the source on both coarse, $4.5^\circ \times 4.5^\circ$, and fine grid, $0.5625^\circ \times 0.5625^\circ$.

In order to enable comparison with the ETEX-II retrievals we show in Fig. 9 a reconstructed profile corresponding to the one given in Fig. 6 and the spatial distribution of the reconstructed mass in Fig. 8, an analogue of Fig. 5.

3.3 Remarks

The localisation of the ETEX-II source is rather good. 71% of the recovered mass is attributed to the correct grid-cell, as compared to 65% in the ETEX-I case. This can be explained by the optimality system to deblur some of the data. It is also explained by very high sensitivity of the source to the nearby measurements which makes the Monterfil grid-cell a very likely spot.
We note that the inverse modelling analysis of ETEX-II signals weak consistency in the data. It can be observed in Fig. 10, where the measurements arising from the reconstructed source are largely scattered as a function of the real measurements. Indeed, the inconsistency could have already been remarked in data - simulation comparison. The inverse modelling analysis, however, goes one step further because it is capable of accounting for errors. Since model is taken into account in this study, and the errors are retrieved by the reconstruction method, the detected inconsistencies go beyond statistical ones which could be diagnosed through geostatistical procedures (Dubois et al., 2005). Still, consistency cannot be improved.

4 Conclusions

We have recalled here the results of source retrieval in the case of the ETEX-I release. The reconstruction performed with the method of maximum entropy has been shown to render satisfactory results in this case. The same procedure employed in source retrieval of the ETEX-II release gives much worse results, although an optimal framework of inversion has been much searched for.

Similarity of both experimental set-ups let us suppose that it is not the inversion method which is to blame. It does perform well in the former case. One could, however, point several other reasons for the difficulties encountered by the inversion procedure in the latter case.

Firstly, the number of the measurements is limited which has an impact on the attainable source resolution (coarser than for ETEX-I). Secondly, the quality of the inversion is highly dependent on the possibility of incorporating error into the modelling procedure. Because convection is not properly taken into account, merely fronts, a sudden splitting of the plume may have been missed, resulting in an important model error and a loss of mass not reproduced in the modelling.

However, this could be ascribed as well to the poor quality or representativeness of the measurements. The official report states that 75% of the mass is unaccounted for.
from the measurements. This could be either due to poor modelling or poor measure-
ments, or both. Our study rises that level up to 85%, as if the observed concentrations
had been diluted by an overall mean factor of 6.8. From the reconstruction point of
view the data seem inconsistent, in particular the observations performed in the early
evolution of the plume (over France and Benelux).

Other clues, independent from this study, plead against the data. Although PMCP
concentrations extracted from the collected samples are of good quality, the samples
themselves are more difficult to judge. In many stations cloud’s presence was intermit-
tent. In particular, several cases are reported in which some stations exhibit cloud pas-
sage twice during the sampling period while another station, located close by, misses
the second peak. Hence, the measurements suggest plume was chopped into several
clouds which moved separately with different velocities (Nodop, 1998). This strange
non-physical behaviour is, however, not supported by any of the models involved in
model intercomparison (Graziani et al., 1998b). Moreover, the aircraft data (Girardi
et al., 1998) do not reveal any splitting of the plume. In addition, even though it has
been argued against any water dilution of the samples, it has also been mentioned that
the instruments’ catalyst (Palladium) have been repeatedly poisoned by an unknown
compound, so that the catalyst needed replacement several times.

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Fig. 1. Plume of the ETEX-II release simulated by POLAIR3D. The units are in ng m$^{-3}$. The snapshots’ dates correspond to 3, 12, 24, 28, 72, and 96 h after the start of the release.
Fig. 2. Duality test. Measurements computed according to the measurement equation, Eq. (1) with $H$ obtained in the dual mode (vertical axis) versus measurements obtained via sampling of a concentration field in a direct simulation (horizontal axis).
Fig. 3. L-curve for a varying standard deviation. Relative entropy for source and errors on the axes refer to $\mathcal{K}_o$ and $\mathcal{K}_e$ in Eq. (3). The point of inflection can only be identified thanks to the graph of curvature (inlay). Spatial resolution is $2.25^\circ \times 2.25^\circ$. 
Fig. 4. L-curve for a varying mass with an inflection point identified as $m=8$ (square). Results obtained for a spatial resolution of $2.25^\circ \times 2.25^\circ$. 
Fig. 5. ETEX-II. Total reconstructed mass in each spatial grid-cell for the resolution of 2.25°×2.25°. The triangles represent the monitoring stations which gathered the measurements used in the inversion. The disc marks the true release site, Monterfil, France. The units are in kg.
Fig. 6. ETEX-II. Profile of the reconstructed source in Monterfil. Total reconstructed mass is equal to 73 kg and the mass reconstructed in Monterfil is 51 kg. Spatial resolution is 2.25°×2.25°.
Fig. 7. ETEX-II. Marginal contributions of each observation to the total information gained in the assimilation $\mathcal{K}_{\sigma,\varepsilon}$ (top figure), and for the information gained on the source alone $\mathcal{K}_{\sigma}$ (bottom figure). The disc radius is proportional to the value of the sensitivity with respect to the measurement. Darker discs correspond to negative values.
Fig. 8. ETEX-I. Total reconstructed mass in each spatial grid-cell for the resolution of 2.25° × 2.25°. The triangles represent the monitoring stations which gathered the measurements used in the inversion. The disc marks the true release site. The units are in kg.
Fig. 9. ETEX-I. Profile of the reconstructed source in Monterfil. Total reconstructed mass is equal to 336 kg and the mass reconstructed in Monterfil is 224 kg. Spatial resolution is $2.25^\circ \times 2.25^\circ$. 
Fig. 10. Posterior measurement consistency test. Measurements (diagnosed PMCP) simulated from the reconstructed source (Fig. 6) as a function of the real measurements (observed PMCP). The dashed lines mark the discrepancy factor equal to 2.