
Radiation source rate estimation through data assimilation of gamma dose rate measurements for operational nuclear emergency response systems

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Abstract: This paper presents an evaluation of an innovative data assimilation method that has been recently developed in NCSR Demokritos for estimating an unknown emission rate of radionuclides in the atmosphere, with real-scale experimental data. The efficient algorithm is based on the assimilation of gamma dose rate measured data in the Lagrangian atmospheric dispersion model DIPCOT and uses variational principles. The DIPCOT model is used in the framework of the nuclear emergency response system (ERS) RODOS. The evaluation is performed by computational simulations of dispersion of Ar-41 that was emitted routinely by the Australian Nuclear Science and Technology Organisation's (ANSTO) previous research reactor, HIFAR, located in Sydney, Australia. In this paper the algorithm is evaluated against a more complicated

case than the others used in previous studies: There was only one monitoring station available each day and the site topography is characterised as moderately complex. Overall the estimated release rate approaches the real one to a very satisfactory degree as revealed by the statistical indicators of errors.

Keywords: radiation source rate estimation; data assimilation; variational method; Lagrangian model; nuclear emergency response.

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1 Introduction

In nuclear power plant accidents that involve release of radionuclides in the atmosphere, the emission rate of radioactive material is usually unknown. During the emergency phase the estimated source term can differ from the true one by the factor of ten or more (US Nuclear Regulatory Commission, 1990). Therefore, improving source rate estimation is of primary importance. A way to assess the release rate is data assimilation (DA) of gamma dose measurements which are typically available around every nuclear power plant. In this respect an innovative computational method has been recently developed in NCSR Demokritos for estimating the unknown emission rate of radionuclides in the atmosphere. The algorithm is based on assimilation of gamma dose rate measured data in the Lagrangian atmospheric dispersion model DIPCOT (Andronopoulos et al., 2009) used in the framework of the nuclear emergency response system (ERS) RODOS (Raskob, 2007) and uses variational principles. The method is described in Tsiouri et al. (2011, 2012). In the latter work (Tsiouri et al., 2012) the method was successfully evaluated against the fluence rate measurements in field experiment of Ar-41 atmospheric dispersion in Mol, Belgium (Drews et al., 2002). In the present work, the method is evaluated against a more complicated case using gamma dose rate measurements from Ar-41 routine releases at the Australian Nuclear Science and Technology Organisation's (ANSTO) previous research reactor, HIFAR, located in Sydney, Australia. The area around the research reactor is characterised by moderately complex topography and spatially varying land cover. The Ar-41 database that is used for the purposes of the study covers various seasons during 2002 to 2003 and includes measured gamma radiation dose rates from four monitoring stations located in a radius of 5 km around the research reactor. There are 16 days of gamma radiation dose measurements but only one monitoring station is available each day. Therefore, the challenge for improving the source rate in this case is the assimilation of gamma dose rate measured data from only one monitoring station and the complex terrain of dissected plateaus and valleys that surrounds ANSTO.

2 Methodology

2.1 Model description

DIPCOT (Andronopoulos et al., 2009) is a three-dimensional model, which simulates atmospheric dispersion estimating particle (puff's) trajectories. It has been comprehensively validated against numerous field and laboratory experiments on atmospheric dispersion (e.g., Andronopoulos and Bartzis, 2010) and it is included in the European real-time, online, decision support (RODOS) system for nuclear emergencies. In DIPCOT there are two modes of particles/puffs movement, the stochastic mode (SM) and the deterministic mode (DM). In DM puffs are transported by the average wind field and grow in size according to well-known Pasquill-type relationships. In SM puffs are transported also by wind fluctuations based on the Langevin equation, formulated for stationary homogeneous isotropic turbulence at the horizontal direction, and on inhomogeneous Gaussian turbulence for the vertical direction, i.e., particles' equations of movement become stochastic. Concentration C (activity concentration of nuclides in air) and gamma dose rates at a particular location and time are calculated by summing the contribution of all neighbouring puffs. A description of the gamma dose calculation methods used in DIPCOT is given in Andronopoulos et al. (2009) and Andronopoulos and Bartzis (2010).

2.2 DA algorithm

An innovative and efficient methodology based on variational DA is used for estimating the unknown emission rate (Tsiouri et al., 2011, 2012). The main objective of the DA method is the minimisation of the following cost function with respect to the control vector $\bar{\psi}$ which consists of the source rates corresponding to times of releases of puffs: \bar{q} .

$$J = J_1 + J_2, \quad J_1 = (\bar{\psi} - \bar{\psi}^b)^T \underline{\underline{B}}^{-1} (\bar{\psi} - \bar{\psi}^b)$$

$$J_2 = \sum_{n=1}^{N_o} \sum_{k=1}^K \sigma_{O^o}^{-2} \left(d_k^o(t_n) - \tilde{d}(\bar{r}^k) \right)^2 = \left(\bar{d}^o - \underline{\underline{G}}\bar{\psi} \right)^T \underline{\underline{O}}^{-1} \left(\bar{d}^o - \underline{\underline{G}}\bar{\psi} \right) \quad (1)$$

Here, $\bar{\psi}^b$ is the first guess estimation of the control vector, $\underline{\underline{O}}$, $\underline{\underline{B}}$ are the covariance matrices of the errors of the observations and the background errors respectively; the vector $\bar{d}^o \in R^{N_o K}$ consists of the gamma dose rates $d^o(n, k)$, measured on each time interval Δt_n by the k^{th} station. The elements of \bar{d}^o are ordered sequentially as follows:

$$d_l^o = d_{(n-1)K+k}^o = d^o(n, k).$$

For substantial improvement in numerical efficiency and accuracy and to enable using the DA method also in the framework of the stochastic Lagrangian atmospheric dispersion models, the control vector reduction technique explained in detail in Tsiouri et al. (2012) is used. This technique is based on the assumption that

during small enough time interval Δt , the source rate can be considered as constant with sufficient accuracy. Then the particles could be joined in $P = N_p / \Pi$ groups with Π particles in each group being characterised by the same source rate: $q_{(j-1)\Pi+1} = q_{(j-1)\Pi+2} = \dots = q_{j\Pi} = \tilde{q}_j$, $1 \leq j \leq P$. Here \tilde{q}_j are the values characterising the source rate of the j^{th} group of particles, which form the reduced control vector: $\underline{\tilde{q}}$ of size P . Clearly, the value of P depends on the choice of the time interval Δt during which the source rate could be considered as constant and thus it is a free variable that depends on the expert judgement of the user. Note that if $P = 1$, then the source rate is assumed to be constant during the whole release interval. Instead of initial problem of minimising equation (1) with respect to the control vector \bar{q} consisting of the release rates of individual particles the ‘reduced’ minimisation problem is solved in which the same function is minimised with respect to the reduced control vector $\underline{\tilde{q}}$. The cost function [equation (1)] with constraint of positive control vector values is minimised using the IMSL ® package (IMSL MATH/LIBRARY, 1987).

3 Applications – results

In the present work, DA runs are performed and the DA algorithm is evaluated against the measured release rate from the Ar-41 database developed at the ANTSO previous research reactor, HIFAR, located in Sydney, Australia. Specifically, 16 different cases are simulated that cover winter and summer periods of the years 2002 and 2003 and include all the atmospheric stability conditions. The Ar-41 database used for the method evaluation include the Ar-41 stack emission rate, measured meteorological data from two stations and measured gamma dose rates from four monitoring stations located in a radius of 5 km around the reactor. All the above data were available in 15-min time intervals. The terrain elevation and the land cover were available on a grid of 25 m resolution for the area of interest around the site. Figure 1 shows the computational domain with terrain elevation contours, the Ar-41 release location, the meteorological stations and the gamma dose rate detectors. The terrain is moderately complicated with hills of about 190 m height and a valley that transverses the domain. The land cover is varying, including urban (south-east part), suburban (central part), woods (along the river) and low vegetation (north and south-west part) areas. The available meteorological measurements included wind speed, wind direction and temperature at the levels of 10 and 49 m for the station Met00 and wind speed and direction at 18.5 m for the station Met01 (Figure 1). The atmospheric stability has been determined by the pre-processors for each 15-min time interval from the temperature gradient between 10 and 49 m and from the wind speed. The raw wind velocity data of 1 min sampling at heights 69 m and 78 m have been averaged on 10-min intervals to drive the dispersion model, together with the Pasquill-Gifford stability categories given in 10-min intervals in the database. The meteorological data were pre-processed by the meteorological pre-processor FILMAKER of the RODOS system to prepare input meteorological fields for the DIPCOT model.

Simulations with the DIPCOT model have been performed with the following set of parameters. The puffs were released at a time interval $\tau \approx 2s$ and at a time interval $\tau \approx 4s$. Simulations have been performed in the SM of DIPCOT operation. The first guess source

emission rate was set by a factor of 10 greater than the true rate. Different number of source time intervals (parameter P of CVR technique) has been used in different runs. A detailed description of the simulations is given in Table 1.

Figure 1 The computational domain with terrain elevation contours, the Ar-41 release location, the meteorological stations (Met00, Met01) and the gamma dose rate detectors (Det 9, Det 16, Det 17, Det 18) (see online version for colours)

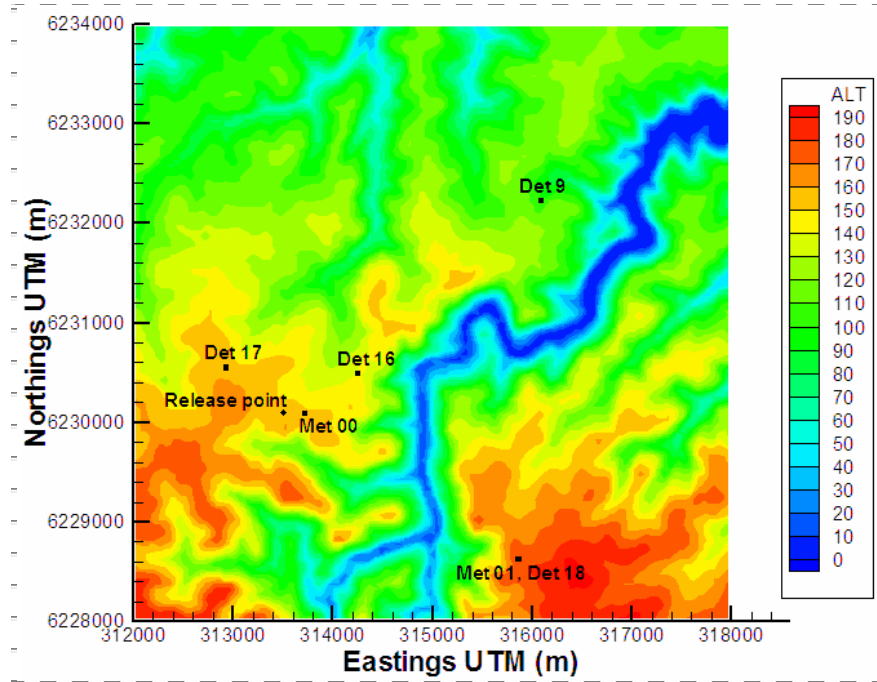


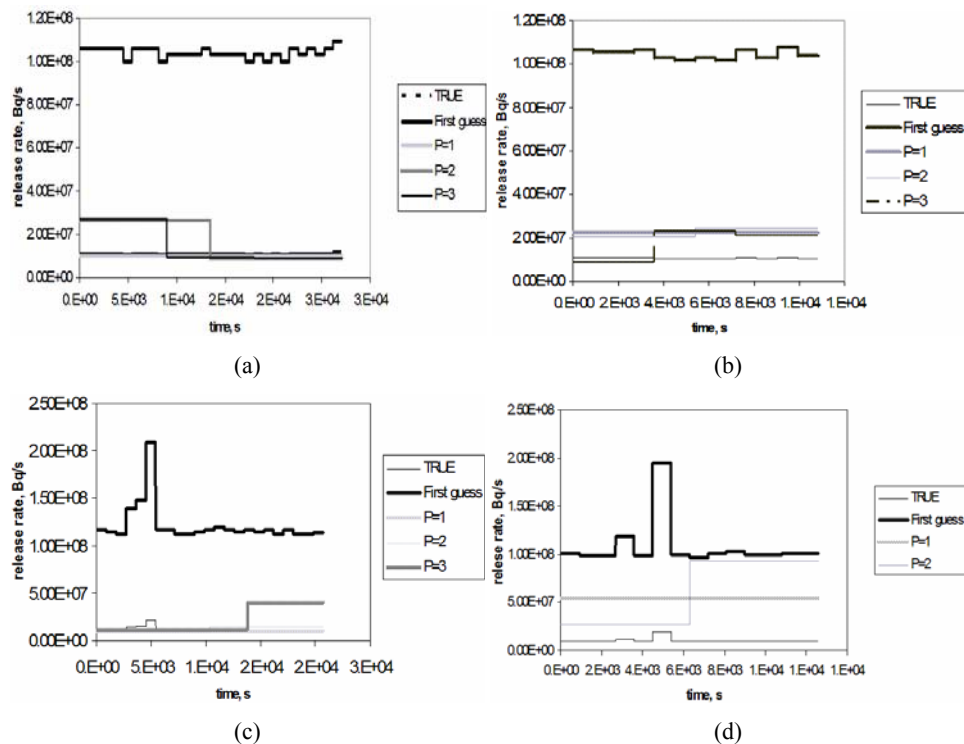
Table 1 The detailed description of the simulations

Case (date)	Station name	Detector point (station #)	Distance from HIFAR (km)	Case details	Tested value of P /corresponding values of time intervals (Δt) during which the source rate is assumed to be constant
Day 1 (06/06/2003)	Waste services (WS)	17	0.73	Winter, unstable conditions	1, 2 and $3/\Delta t \sim 450, 225, 150$ min
Day 2 (17/12/2002)	Main gate (MG)	16	0.82	Summer, stable conditions	1, 2 and $3/\Delta t \sim 180, 60, 30$ min
Day 3 (22/06/2003)	Boys town (BT)	18	2.78	Winter, stable conditions	1, 2 and $3/\Delta t \sim 405, 202, 5, 135$ min
Day 4 (09/07/2003)	Barden ridge (BR)	9	3.33	Winter, stable conditions	1, 2 and $3/\Delta t \sim 210, 105, 70$ min

Figure 2 shows how efficient are the DIPCOT model in estimating the unknown source rate with the implementation of the DA algorithm. This figure presents the source emission rate estimations as result of the assimilation of gamma dose rate data for the

cases of day 1, day 2, day 3 and day 4 in case of the stochastic version of DIPCOT against time. The time in x-axis is the time of release. The puffs were released at a time interval $\tau \approx 2s$. The source rate is estimated at fixed intervals over the release period of time depending on the tested value of P . Results with different number of groups P are presented. The true source rate is known but we supposed we do not know it and an arbitrary value of it was taken (ten times more). The 'first guess' source rate and the true source rate are not constant (first guess is ten times more than true). The other lines represent the estimated source rate for different values of P . As already explained in the section of methodology the value of P depends on the choice of the time interval Δt during which source rate is considered as constant (if $P = 1$, then the source rate is assumed to be constant during the whole release interval).

Figure 2 Release rate estimations as result of gamma dose rate assimilation, for (a) day 1 – WS station (Det 17), (b) day 2 – MG station (Det 16), (c) day 3 – BT station (Det 18) and (d) day 4 – BR station (Det 19) cases



As it can be clearly seen from Figure 2 the algorithm managed to estimate the unknown source rate. The adjusted source functions in all cases are much better than the first guess source function. Overall the estimated release rate approaches the real one to a very satisfactory degree for day 1 (Det 17), day 2 (Det 16) and day 3 (Det 18) cases under all stability conditions. For day 4 (Det 9) because of the small sample size the results are less satisfactory, but even in this worst case the algorithms improves the source rate.

It is important to mention that the complexity of each case tested is increased by the complexity of terrain. The complexity of terrain results in more difficult accurate calculations in forward run of the model and consequently it increases difficulty of

assimilation problem. Additionally, the difficulty of assimilation problem is increased as the number of available monitoring stations is decreased. In the present work, one monitoring station was available each day and the site topography is characterised as moderately complex. Therefore, even with assimilation of gamma dose rate measured data from only one monitoring station and moderately complex site topography the DA method allows for substantial improvement of source rate.

Table 2 Mean absolute relative error (MAE) and mean relative biases (MRB) of calculated source function as compared to measured source function

Case	Station name	Detector point station #	No. of puffs	P	MAE	MRB
Day 1	Waste services (WS)	17	13,500	First guess	9.0	9.0
Day 1	Waste services (WS)	17	13,500	3	0.62	0.46
Day 1	Waste services (WS)	17	13,500	2	0.86	0.71
Day 1	Waste services (WS)	17	13,500	1	0.06	-0.06
Day 1	Waste services (WS)	17	6,300	First guess	9.0	9.0
Day 1	Waste services (WS)	17	6,300	3	1.3	1.14
Day 1	Waste services (WS)	17	6,300	2	1.12	1.02
Day 1	Waste services (WS)	17	6,300	1	0.02	-0.01
Day 2	Main gate (MG)	16	5,400	First guess	9.0	9.0
Day 2	Main gate (MG)	16	5,400	3	0.80	-0.67
Day 2	Main gate (MG)	16	5,400	2	1.12	1.12
Day 2	Main gate (MG)	16	5,400	1	1.12	1.12
Day 3	Boys town (BT)	18	10,344	First guess	9.0	9.0
Day 3	Boys town (BT)	18	10,344	3	0.88	0.70
Day 3	Boys town (BT)	18	10,344	2	0.28	-0.09
Day 3	Boys town (BT)	18	10,344	1	0.17	-0.17
Day 4	Barden ridge (BR)	9	6,300	First guess	9.0	9.0
Day 4	Barden ridge (BR)	9	6,300	2	4.58	4.58
Day 4	Barden ridge (BR)	9	6,300	1	4.11	4.11

Note: Errors of the first guess source function as well as the errors of source functions corrected in assimilation runs with different values of CVR parameter P .

This qualitative result is confirmed with the results of the mean relative absolute error (MAE) and the mean relative bias (MRB) presented in the Table 2 ($MAE = \langle |q^a - q^t| \rangle / \langle q^t \rangle$, $MRB = \langle q^a - q^t \rangle / \langle q^t \rangle$), where q is the source function, $\langle \rangle$ means averaging, superscripts 'a' and 't' denote the analysed and the true source function respectively). The results obtained by the stochastic version of DIPLOT at a time interval $\tau \approx 2s$ for day 1, day 2, day 3 and day 4 cases and with different values of the CVR parameter P are presented in the Table 2. The results by setting the time interval that the puffs were released to $\tau \approx 4s$ are also presented for the day 1. Generally, as follows from these results in all cases the analysed source rate in the assimilation runs is much better than the first guess function even if in the forward run (e.g., day 4) the model did not succeed in attaining the suggested satisfactory performance as reported in Andronopoulos

et al. (2010). Satisfactory results also obtained even if we reduce the no. of puffs to half as it can be easily seen in Table 2 (for day 1 case).

4 Conclusions

The innovative and efficient DA method that has been recently developed in NCSR Demokritos for estimating an unknown emission rate of radionuclides in the atmosphere is evaluated using gamma dose rate measurements from Ar-41 routine releases at the ANTISO previous research reactor, HIFAR, located in Sydney, Australia. The area around the research reactor is characterised by moderately complex topography and spatially varying land cover. The Ar-41 database that is used for the purposes of the study covers various seasons during 2002 to 2003. The method is based on the assimilation of gamma dose rate measured data in the Lagrangian stochastic atmospheric dispersion model DIPCOT and uses variational principles. The DIPCOT model is used in the framework of the nuclear ERS RODOS. In the DA runs performed in this study, the first guess source emission rate has been set by a factor of 10 greater than the true one. In all cases of DA runs the statistical indicators of errors of the estimated source emission rate as compared to the measured one were reduced. In all cases the estimated release rate approaches the real one to a very satisfactory degree as revealed by the statistical indicators of errors under all stability conditions. Even for the day 4 (Det 9) with small sample size the algorithm improves the source rate. The DA method is successfully evaluated against a complicated case, under a range of atmospheric stability conditions. There was only one monitoring station available each day, therefore even with assimilation of gamma dose rate measured data from only one monitoring station the DA method allows for substantial improvement of source rate. Therefore, the presented results demonstrate the potential of the developed DA algorithm for application in operational nuclear ERSs.

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