Source apportionment using LOTOS-EUROS: module description and evaluation

R. Kranenburg, C. Hendriks, M. Schaap, and A. Segers

TNO, P.O. Box 80015, 3508 TA, Utrecht, The Netherlands

Received: 30 October 2012 – Accepted: 9 November 2012 – Published: 28 November 2012

Correspondence to: R. Kranenburg (richard.kranenburg@tno.nl)

Published by Copernicus Publications on behalf of the European Geosciences Union.
Abstract

To design effective mitigation strategies the origin of air pollutants needs to be known. Chemistry transport models can be used to assess the origin of air pollution across a large domain. However, in traditional simulations the information on origin is lost and brute force scenario studies are performed to assess the origin. Alternatively, one can trace the origin of air pollutants throughout a simulation using a labeling approach. In this paper we document and demonstrate a newly developed labeling module for the chemistry transport model LOTOS-EUROS which tracks the source allocation for all particulate matter components and precursor gases. Dedicated simulations confirmed that the new module functions correctly. The new module provides more accurate information about the source contributions than using a brute force approach with scenario runs as the chemical regime remains unchanged. An important advantage of the new module is the reduction of computation costs and analysis work associated with the calculations. The new module was applied to assess the origin of particulate nitrate across the Netherlands. Averaged across the Dutch territory the main contributions to nitrate derive from road and non-road transport as well as power plants. Overall, only one-fifth of the concentration derived from sources located inside the country. The new technology enables new research directions as improved information on pollution origin is desired for policy support as well as scientific applications.

1 Introduction

Exposure to particulate matter (PM) in ambient air leads to human health problems (Dockery et al., 1993; Klemm et al., 2000). Deposition of secondary inorganic aerosol and its precursors leads to a loss of biodiversity through acidification and eutrophication of soils and surface waters (e.g. Bobbink et al., 2010). Moreover, particulate matter components play a key role in climate change affecting the radiation balance of the earth (Forster et al., 2007). To limit the effects of PM pollution, efforts are made to
reduce emissions of PM and its precursors. To design cost effective mitigation strategies, a thorough understanding of the sources of particulate matter is crucial. As PM consists of a host of components with different sources and atmospheric behaviour, establishing the origin of PM remains a challenge. Detailed field campaigns that are designed to establish the origin of PM differ in complexity. Composition and tracer data have often been used to interpret PM time series and origin. Chemical mass closure studies are reported for many regions (Putaud et al., 2010). These detailed chemical speciation data sets enable the use of statistical approaches, such as Positive Matrix Factorisation, to identify PM origin (Viana et al., 2011). However, these methods are only able to distinguish between a limited number of broad source categories. Furthermore, they are typically not able to provide a source apportionment for secondary components. Hence, additional data are needed to establish a full source apportionment.

Complementary to experimental data, a chemical transport model (CTM) can be used to obtain a more detailed source apportionment. Chemical transport models provide calculations of the evolution of the air pollution situation across a region based on emission inventories and atmospheric process descriptions. Hence, they implicitly contain the information to perform a source apportionment. Numerous studies have employed a brute force approach to gain insight in source contributions. In these studies scenario runs are used reducing the emissions of sources under investigation and comparing the differences to the base case (e.g. Lane et al., 2007). Although the computational burden is large, this approach provides a good insight for of inert species. However, in case of species involved in atmospheric chemistry, a negative impact on the source apportionment results may occur as perturbing emissions may cause nonlinear effects.

Presently, few models are equipped with modules to overcome these negative impacts on chemical reactive species. Most divide the PM components into source-specific species that are tracked separately through the model (e.g. McHenry et al., 1992; Ying and Kleeman, 2006). These approaches are capable of accurately track-
ing the source contribution for secondary species. However, the computational burden remains as large as using a scenario approach limiting their use for a large number of sources. Yarwood et al. (2007) and Wagstrom et al. (2008) present the Particulate Source Apportionment Technology (PSAT) algorithm within CAMx, combining the capability of accurately dealing with secondary species with limited CPU demand. With PSAT the concentration of the species is modelled as before, but next to this the fractional contribution of all sources is kept track off through all processes. CAMx incorporates detailed process descriptions, but requires a large computational time to perform simulations over long time periods. Hence, applications to the full European domain do not exist. We developed a source apportionment model for the operational CTM LOTOS-EUROS (Schaap et al., 2008) to be able to study the origin of particulate matter in Europe in more detail. In this paper we document, validate and demonstrate the new source apportionment module inspired on the PSAT approach.

We apply the source apportionment system to study the origin of ammonium nitrate in the Netherlands. Ammonium nitrate is the most important component of particulate matter in the Netherlands (Weijers et al., 2011). In the Netherlands ammonium nitrate levels rise more than proportionally with total PM (Weijers et al., 2011), which has also been observed in other countries (Putaud et al., 2010). We focus on nitrate as it is a product of complex chemistry and its precursor originates from multiple source sectors. The source apportionment for nitrate is part of a larger effort to assess the origin of PM in the Netherlands as reported by Hendriks et al. (2012).

In this paper we first present a short overview of the LOTOS-EUROS model (Chapter 2). Second, a detailed description of the labeling source apportionment technique is given (Chapter 3). Next, the results of a technical validation to ensure a proper functioning of the system for the different processes are presented (Chapter 4). Also, the application aimed to establish the origin of nitrate in the Netherlands is shown (Chapter 5). The paper is concluded with a summary of the main results and a short discussion on the benefits of the developed module.
2 The LOTOS-EUROS chemistry transport model

The model employed in this study is the 3D chemistry transport model LOTOS-EUROS, which is aimed at the simulation of air pollution in the lower troposphere. The model is of intermediate complexity in the sense that the relevant processes are parameterized in such a way that the computational demands are modest enabling hour-by-hour calculations over extended periods of several years within acceptable CPU time.

The current master domain of LOTOS–EUROS is bound at 35° and 70° North and 30° West and 60° East. The model projection is normal longitude–latitude and the standard grid resolution is 0.50° longitude × 0.25° latitude, approximately 25 × 25 km². The actual domain for a simulation can be set as long as it falls within the master domain as specified above. In addition, it is possible to increase or decrease the resolution up to a factor 4. In the vertical, the model extend to 3.5 km above sea level and uses the dynamic mixing layer approach to determine the model vertical structure. This means that on top of a 25 m surface layer a well-mixed boundary layer is assumed. The height of the mixing layer is obtained from the ECMWF meteorological input data used to drive the model. The height of the reservoir layers is determined by the difference between ceiling (3.5 km) and mixing layer height. Both layers are equally thick with a minimum of 50 m. In a few cases, when the mixing layer extends near or above 3500 m, the top of the model exceeds the 3500 m according to the above-mentioned description.

The advection in all directions is handled with a monotonic advection scheme (Walcek et al., 1998). Gas phase chemistry is described using the TNO CBM-IV scheme, which is a condensed version of the original scheme (Whitten et al., 1980). Hydrolysis of N₂O₅ is described explicitly (Schaap et al., 2004). Cloud chemistry is described following Banzhaf et al. (2012). Aerosol chemistry is represented using ISORROPIA2 (Nenes et al., 2007). The dry deposition in LOTOS-EUROS is parameterized following the well-known resistance approach following the EDACS system (Erisman et al., 1994), including a compensation point approach for ammonia (Wichink Kruit et al., 2012). The aerodynamic resistance is calculated for all land use types separately.
Below cloud scavenging is described using simple scavenging coefficients for gases (Schaap et al., 2004) and particles (Simpson et al., 2003).

Anthropogenic emissions are prescribed using the TNO-MACC emission inventory (Kuenen et al., 2011). The temporal variation of the emissions is represented by time factors. For each source category a monthly factor breaks down the annual total into monthly totals. This value is multiplied by a factor for the day of the week (i.e. Monday, Tuesday etc.) and finally with a factor for the hour of the day (local time). The LOTOS-EUROS model includes a biogenic emission routine based on detailed information on tree species over Europe (Koeble and Seufert, 2001). The emission algorithm is described in Schaap et al. (2009) and is very similar to the simultaneously developed routine by Steinbrecher et al. (2009). Sea salt emissions are described using Martensson et al. (2003) and Monahan et al. (1986) for the fine mode and coarse mode, respectively.

The LOTOS-EUROS model has participated in several international model intercomparison studies addressing ozone (Hass et al., 1997; Van Loon et al., 2007; Solazzo et al., 2012a) and particulate matter (Cuvelier et al., 2007; Hass et al., 2003; Stern et al., 2008; Solazzo et al., 2012b) and shows comparable performance to other European models.

3 Source apportionment module

3.1 Overview

To track the origin of the modeled concentrations for different tracers, a source apportionment module for LOTOS-EUROS was developed. Using a labeling technique the new modules calculates the contribution of specified sources for all model grid cells and time steps. In this calculation the contributions per label are calculated as fractions of the total tracer concentration. As the fractions must add up to one for mass conservation, all processes that are sources and sinks of mass within the model must
be accounted for in the source apportionment module, including initial and boundary conditions. The calculations to track the source contributions differ per process; the processes in the model can be categorized in four groups:

- Emissions
  - Anthropogenic emissions
  - Natural emissions
- Transport processes
  - Advection
  - Diffusion
  - Adjust
  - Sedimentation
- Removal processes
  - Dry deposition
  - Wet deposition
- Chemistry
  - Gas phase chemistry
  - Aerosol chemistry

For each of these processes the calculations are described below.
### 3.2 Emissions

The specification of the labels to track throughout the model simulations is done in the emission routine. In principle, the definition of the labels is very flexible and any kind of allocation is possible as long as the required detail is present in the input data. As one is mainly interested in the source contribution of the anthropogenic sources, the natural emissions (sea salt, biogenic NMVOC and windblown dust) are treated separately and obtain a separate label.

For each hour in the simulations the base emission data containing the annual average emission per sector and country for each grid cell is processed and added into the emission array using the sector dependent time profiles. During this processing the source contributions are also accounted in the array $fr_{emis}(l)$. During each time step the concentration changes due to all emissions ($c_f$) are added to the existing concentrations $c_0$ and the new source allocation $fr_n(l)$ is updated by a straight forward weighted average of existing and concentration change:

$$fr_n(l) = \frac{fr_0 \cdot c_0 + fr_{emis} \cdot c_f}{c_n}$$

In the equation above $fr_n(l), fr_0(l)$ are the new and the old fractions for label $l$, while $fr_{emis}(l)$ is the fraction of the emission for label $l$. The total old and new concentrations are given by $c_0$ and $c_n$, while the emitted mass flux converted to a concentration flux is given by $c_f$.

### 3.3 Transport processes

In the advection scheme (Walcek et al., 1998) the advection operator is performed sequentially in three dimensions. The concentration fluxes through the cell edges are calculated based on the fluxes between two nearest upwind and the nearest downwind
cell. The fluxes are limited by the maximum and minimum concentration of the two surrounding cells. Due to this application, the advection process is slightly non-linear.

For the update of the source allocation, the mass fluxes between all cells are passed to the labeling routine. The source allocation of the influx is taken as the source allocation of the nearest upwind cell. The new source allocation in each cell is then calculated as a weighted average of the remaining source allocation and the source allocations of the influx(es) of the cell.

For the update of the source allocation, the mass fluxes between all cells are passed to the labeling routine. The source allocation of the influx is taken as the source allocation of the nearest upwind cell. The new source allocation in each cell is then calculated as a weighted average of the remaining source allocation and the source allocations of the influx(es) of the cell.

\[ fr_n(l) = \frac{fr_o(l) \cdot c_o + F(l)}{V} \]

Here, \( F(l) \) is the mass flux for label \( l \), while \( V \) is the volume of the cell, \( c_o \) and \( c_n \) are the old and new concentration. The mass flux per label is defined as the influx minus the outflux.

\[ F(l) = fr_o(l)_{in} \cdot F_{in} - fr_o(l)_{out} \cdot F_{out} \]

The fractions \( fr_o(l)_{in} \) for the mass influx \( F_{in} \) are taken from the donor cell, while the fractions \( fr_o(l)_{out} \) for the mass out flux \( F_{out} \) are taken from the cell itself.

All other processes that govern one-dimensional transport, i.e. diffusion, sedimentation and adjust are implemented following the same approach. The process adjust accounts for the concentration changes due to the adaption of the vertical layering structure as a consequence of the rise and fall of the mixing layer.
3.4 Wet and dry deposition

Wet and dry deposition are important sinks in the model. As it is assumed that each molecule of a species has the same probability to be deposited, the source apportionment for the concentration arrays does not change. A complication is associated with the treatment of the re-emission of ammonia which is possible as the model includes the bi-directional surface-atmosphere exchange (Wichink Kruit et al., 2011). Here we assume that the exchange is fast and the source allocation of this emitted ammonia is given the same allocation as the concentration in air. Alternatively, one could store the source allocation of the ammonia deposition fluxes and use these for the re-emission flux. Given the short atmospheric lifetime and the dominant impact of agriculture for ammonia we feel that both approaches would yield very similar results.

3.5 Gas phase chemistry

The gas phase chemistry in LOTOS-EUROS is described chemistry using a CBM4 mechanism. We aim to provide a source attribution that is valid for the current atmospheric conditions. As a consequence, the labeling process is only implemented for chemical active tracers with an N, C or an S atom. These atoms are conserved and are traceable. This means that four oxidants (OH, H₂O₂, HO₂ and O₃) and two operator species (XO₂ and XO₂N) are not traced. The following differential equation is solved to calculate the new concentrations of all species.

\[ \frac{dc}{dt} = c_p + c_l c(t) \]

Here, \( c_l \) contains the loss rates per time step (\( dt \)), \( c_p \) contains the production rates per time step and \( c \) represents the concentrations of all species. This system is solved with a numerical implicit method, with the following result:

\[ c_n = \frac{c_0 + dt \cdot c_p}{1 + dt \cdot c_l} \]
In here \(c_o\) and \(c_n\) are the new and old concentrations for each species.

The main assumption for the labeling for the chemistry is that first the species are produced before their loss is taken into account. For the source allocation of each species it is necessary to keep track of the produced material as function of its originating species containing the corresponding C, N or S atom. One of the main advantages in the chemical scheme used in LOTOS-EUROS is that this information is easily available. From the solver the production rates split into origin of species are passed to the labeling routine using a square production matrix \((Y_p)\) with the size of the number of labeled species. The columns indicate the reacted species, while the rows contain the produced species. For example: the entry with column index of NO and row index of \(NO_2\) corresponds with the produced \(NO_2\) out of NO, which is a composition of all production rates which deals with production of \(NO_2\) out of NO. In the labeling routine the new source contributions are calculated with the following formula:

\[
fr_n(l) = \frac{fr_o(l) \cdot c_o(l) + dt \cdot fr_o(l) \cdot Y_p}{c_o + dt \cdot c_p}
\]

Here, \(fr_o(l)\) and \(fr_n(l)\) are the old and new fractions of label \(l\), while \(c_o(l)\) is the old concentration for label \(l\). Note that the first multiplication in the numerator is a vector-vector multiplication, which multiplies the entries of both vectors one by one. The second multiplication is a full matrix-vector multiplication, which are generally computationally expensive. As the matrix \(Y_p\) is sparse, a special routine for sparse matrices is used to save computer time.

To illustrate the process we use the reaction of nitrogen dioxide with the hydroxyl radical as an example:

\[
NO_2 + OH \rightarrow HNO_3 \text{ with rate } k.
\]

This reaction produces nitric acid (\(HNO_3\)) and results in a loss of nitrogen dioxide (\(NO_2\)). The reaction clearly illustrates that the N-atom is conserved. This reaction will lower the \(NO_2\) concentration but will not change the source apportionment of \(NO_2\) as all \(NO_2\) molecules have the same chance to react. The produced \(HNO_3\) has the origin of...
NO₂. The new concentration per label for HNO₃ \(c_n(\text{HNO}_3, l)\) is calculated as follows (only the reaction above is involved):

\[
fr_n(\text{HNO}_3, l) = \frac{c_o(\text{HNO}_3) \cdot fr_o(\text{HNO}_3, l) + Y_p(\text{HNO}_3, \text{NO}_2) \cdot fr_o(\text{NO}_2, l)}{c_o(\text{HNO}_3) + c_p(\text{HNO}_3)}
\]

With \(Y_p(\text{HNO}_3, \text{NO}_2) = k \cdot [\text{NO}_2] \cdot [\text{OH}]\). The term \(Y_p(\text{HNO}_3, \text{NO}_2) \cdot fr_o(\text{NO}_2, l)\) corresponds with the label specific production of HNO₃ from NO₂. The term \(c_p(\text{HNO}_3)\) corresponds with the total produced HNO₃. In the case with only one reaction, this term is equal to \(Y_p(\text{HNO}_3, \text{NO}_2)\).

In the current implementation it is assumed that for the production of species with C, N or S atoms the source apportionment is taken from the original species with the corresponding atom. This means that for reactions between organic compounds and nitrogen oxides such as the oxidation of formaldehyde by the NO₃ radical the origin of the produced HNO₃ is taken from the that of NO₃ while the origin of the produced CO is taken from that of formaldehyde. The reactions forming and dissociating Peroxy Acetyl Nitrate (PAN) are an exception to this rule as PAN combines both C and N atoms in a single molecule. It is an important reservoir species and may be transported over quite a distance. At some point PAN will dissociate, but within the current implementation there is no way to keep track of the original N or C source contributions. Therefore, the source apportionment of the produced PAN is taken from both precursor species (\(\text{C}_2\text{O}_3\) and \(\text{NO}_2\)). For the dissociation reaction the normal procedure is followed. Consequently, this is the only way for sources without an N-emission to receive a small but non-zero contribution to the N-species. Note that the current implementation enables to exclude the labeling of the C-atoms by limiting the \(Y_p\) matrix to the inorganic chemistry only. This feature was implemented as an option as it gives a slightly more accurate source allocation for the inorganic species and, more importantly, reduces computing time since the matrix vector multiplication is much smaller. In this study the full source apportionment is used.
3.6 Aerosol chemistry

The source apportionment of non-volatile aerosol components such as sulfate and sodium nitrate is straightforward and follows the approach outlined for the gas phase chemistry. However, a different approach is taken for a semi-volatile component such as ammonium nitrate. The thermodynamic equilibrium module ISORROPIA-II is used in LOTOS-EUROS to assess the partitioning between nitric acid (HNO₃) and nitrate (NO₃), as well as between ammonia (NH₃) and the ammonium (NH₄). In a thermodynamic equilibrium the exchange between the gas to aerosol phase is by definition equal. Though no net mass is exchanged, the constant exchange between the two phases will cause that the origin of the gas and aerosol phase will equalize. Hence, after the equilibrium calculation the source attribution of nitrate (ammonium) and nitric acid (ammonia) are averaged:

\[
fr_n(\text{HNO}_3) = fr_n(\text{NO}_3) = \frac{fr_a(\text{HNO}_3)c_o(\text{HNO}_3) + fr_a(\text{NO}_3)c_o(\text{NO}_3)}{c_o(\text{HNO}_3) + c_o(\text{NO}_3)}
\]

\[
fr_n(\text{NH}_3) = fr_n(\text{NH}_4) = \frac{fr_a(\text{NH}_3)c_o(\text{NH}_3) + fr_a(\text{NH}_4)c_o(\text{NH}_4)}{c_o(\text{NH}_3) + c_o(\text{NH}_4)}
\]

4 Technical validation

The development and implementation of the labeling technique was performed for each process at a time. The correct functioning of the code was evaluated with dedicated experiments. For the emission, vertical diffusion, deposition and adjust processes a simple experiment could be designed as their solution is refined to a single column and the processes are linear. For advection and chemistry more elaborate and indirect experiments were performed. Below we present the performance evaluation for those three cases. For the chemistry several species are discussed to illustrate the functioning of the module.
4.1 Case 1: linear processes

To test the labeling technique for the linear processes emission, deposition, diffusion, sedimentation and adjust, the LOTOS-EUROS model was run for the month January 2007, for primary PM components. These include:

- EC fine: Elemental Carbon in fine mode (EC$_f$)
- PPM fine: Primary Particulate Matter in fine mode (PPM$_f$)
- PPM coarse: Primary Particulate Matter in coarse mode (PPM$_c$)

The EC$_f$ emissions are estimated as a percentage of the fine mode primary PM (PPM$_f$) emissions using sector dependent factors (Schaap et al., 2004). This approach enabled to define a simulation in which the EC$_f$ concentration is calculated as a separate tracer and compared to a labeled EC$_f$ fraction of the ppm$_f$ concentration. Thus, within this simulation the EC$_f$ fraction of all primary PM components was labeled using five labels:

- Fraction ec$_f$
- Fraction not ec$_f$
- Boundary conditions (north, east, south and west boundaries)
- Initial condition
- Aloft boundary condition (boundary on top of the domain)

As the boundary and initial conditions were set to zero, their fraction should remain zero. Moreover, the fractions of EC$_f$ for components EC$_f$ and PPM$_c$ should remain 1 and 0, respectively, which they did. The contribution of EC in PPM$_f$ should equal the EC$_f$ concentration in the model. In Fig. 1, the results are shown for the combination of the linear processes in the model. The results shows that the fractions of label EC$_f$ for PPM$_f$ fits perfectly on the ratio between EC$_f$ and PPM$_f$. There are small deviations...
which are caused by the numerical precision in the computations. Tests using all linear processes separately yields the same results. From these experiments it is clear that the source allocation routine for these processes functions correctly.

4.2 Case 2: advection

In the second performance test we add the advection process to the system. We perform an annual simulation for 2007 for a small domain centered around the Netherlands including The Netherlands, Belgium, northern France and the western part of Germany. Simulations were performed for the full year of 2007. The labels were defined to represent the geographical areas:

- Belgian emissions
- German emissions
- Dutch emissions
- Emissions from other countries or seas
- Boundary conditions
- Initial conditions
- Aloft boundary condition

The results of this labeling simulation were compared to four scenario simulations in which only the separate countries emit their emissions. As the primary species are treated to be inert, the system is linear except for the advection and a close resemblance is expected.

In Fig. 2 we present the results for the Dutch contribution to the modeled EC concentration. The annual mean EC distribution for the scenario run containing Dutch emissions (upper left) closely resembles that of the Dutch labeled concentrations for
the full model simulation (upper right). The differences between the estimates for the Dutch EC contribution differs at most 4% (lower panel). These small differences are induced by the Walcek advection algorithm in which the fluxes through the cell edges are calculated as a weighted average of one downwind and two upwind cells. In case of scenario simulations with different concentrations gradients the fluxes through the cell edges will slightly differ between the simulations. This becomes mostly visible around large source areas. Also the impact of the border is visible in the difference due to the steep concentration gradient across the border in case of a scenario run. In a labeling simulation all sources are included and the impact of differences in gradients are avoided. As such, one could argue that the labeling technique provides a more consistent approach than scenario simulations.

4.3 Case 3: chemistry

In principle, it is impossible to validate the functioning of the labeling routine for a full chemistry simulation as the chemistry scheme is non-linear. Comparing to a scenario simulation will show the impact of indirect effects through changes in oxidant levels and the change in pollutant levels. To illustrate the functioning of the module we compare a simulation in which we label a relatively small fraction (5%) of the Dutch emissions to a scenario simulation with the same emission reduction on all anthropogenic sources. The simulations were performed for July 2006 with active photochemistry. Results are discussed for three components with a different origin and lifetime. We first compare the results for carbon monoxide (CO) as its chemistry is relatively simple, it has a long life time and has strong anthropogenic sources. Hence, the impact of changes in chemical regime is expected to be small. Next, we compare the results of a reaction product from the degradation of non-methane volatile organic compounds (NMVOC) being Methylglyoxal (MGLY). This species has a short life time and is largely determined by oxidation of biogenic VOC. Hence, the impact of the chemical regime is expected to be very large. Finally, we discuss NO and NO₂ as they behave typically due to the impact of the photo-stationary equilibrium.
In Fig. 3 the labeled 5% contribution (upper left) is compared to the impact of the 5% emission reduction (upper right). Both simulations produce a contribution of about 2 ppb in the Randstad, with lower values in the more rural part of the Netherlands and highest values in industrialized areas. The difference between both results (Figure 3 lower panel) is an order of magnitude lower than the estimated 5% contribution itself. The non-linearity in the chemistry (and the transport) provides a difference of about 2–10% in the estimated Dutch contribution, which we consider to be relatively small. As the Netherlands is characterized by very high NO\textsubscript{x} emissions, the difference can be explained by a net increase in oxidant levels in the scenario simulation due to the decreasing impact of ozone titration.

In Fig. 4 the results are shown for MGL\textsubscript{Y}, a reaction product of the degradation of NMVOCs. Comparison between the labeled contribution and the scenario difference shows a large difference. The labeled 5% Dutch contribution is a factor 3 higher than the concentration difference due to a 5% reduction in all Dutch emissions. In addition, the patterns are not all the same. We expect that the change in chemical regime impacting all formation and destruction routes has a as large impact on the estimated contribution based on the scenario. To validate this assumption a second simulation was performed with only a 5% reduction on the Dutch NMVOC emissions, keeping all other emissions the same. The lower right panel of Fig. 4 shows the concentration change for MGL\textsubscript{Y} for this simulation. The results are much closer to the labeled simulation, both in magnitude and gradients. Differences are now at most 10% and occur in regions where anthropogenic emissions are largest and non-linearity in the chemistry and its impact on the solver is expected to be largest. The close agreement provides confidence in the functioning of the source apportionment module. These results also clearly illustrate the difference between source apportionment based on scenarios versus a labeling technique.

Considering the photo stationary equilibrium between NO and NO\textsubscript{2} a difference in results is also anticipated for results of these components. NO\textsubscript{x} is largely emitted in the form of NO. NO reacts rapidly with ozone to form NO\textsubscript{2}. The photolysis reaction of NO\textsubscript{2}
during daylight establishes an equilibrium during the day, though NO concentrations in the presence of ozone are only a fraction of that of NO$_2$. Due to the rapid conversion during the day the source apportionment of NO and NO$_2$ will be equal. In Fig. 5 the relative reduction of the NO$_2$ and NO concentrations are presented for the simulation with 5% reduced Dutch emissions. These figures show that the impact on NO$_2$ is a little less than 5% while the impact on NO is mostly larger than 5%. The explanation lies in the difference in the extend of titration between the two runs. Consider a box with a certain amount of ozone. In case this amount has been fully titrated, a reduction in NO emissions may not impact NO$_2$ but will impact NO to a larger extend. This feature is clearly visible in the scenario results. The case shows that the first reduction on NO$_x$ emissions will not be most effective to reduce NO$_2$ levels evidenced by the lower than 5% impact on NO$_2$. However, in a mixture of emission sources all contributions have the same probability to react. The labeling simulation therefore yields much closer values to 5%, which is the actual anticipated result. Again, these results show the functioning of the module and the advantage it may have on a brute force study based on scenario studies.

### 4.4 Computational cost

Here we compare the computational cost of a labeling simulation with $n$ sectors to a brute force source apportionment effort containing $n$ scenarios. The computational time of single simulation using the source apportionment tool is larger than for a single scenario due to the additional bookkeeping and calculations. However, it is only a fraction of the total of all scenario runs. In Fig. 7 the computation time using the new tool as a fraction of the computation time for $n$ scenario runs is given for both simulations with inert and chemically active components. The graph shows that the computer time saved increases with the number of labels. For example, with $n=24$ labels and inert tracers one needs only about 17 percent of the computational time compared to 24 scenario runs. In case the full chemistry is added, the profit is somewhat smaller as the chemistry is the most expensive part in LOTOS-EUROS. Still, the computational costs
with 24 labels are about a factor 4 lower than using scenario runs. Hence, a significant reduction in computational cost is obtained.

5 Application to the origin of nitrate across the Netherlands

To demonstrate the potential applications of the source apportionment module we provide an assessment of the origin of particulate nitrate across the Netherlands. The majority of the nitrate in the Netherlands is present in the form of ammonium nitrate, whereas coarse mode (sodium) nitrate contributes a minor part (Weijers et al., 2011; ten Brink et al., 1997). Both formation mechanisms are present in the LOTOS-EUROS model. Annual mean modeled concentrations underestimate the observed nitrate concentrations by about 20%. Evaluation against daily particulate matter samples (Hendriks et al., 2012) as well as hourly observations (Schaap et al., 2011) show that the model is able to capture the variability in time quite well. For a detailed discussion on the model performance and associated uncertainties we refer to these papers. Here, we present the source apportionment for nitrate obtained with a simulation in which labels were defined to track the national and foreign contributions of all SNAP level 1 emission sectors (20 in total). Note that the interpretation of the results was also made for all PM components and is presented in detail in Hendriks et al. (2012).

Averaged across the Dutch territory the main contributions to nitrate derive from road transport (30%), other transport (25%) and power plants (15%). Overall, only 20% of the concentration derived from sources located inside the country, which is explained by the secondary nature of nitrate, the relatively small size of the country and peak nitrate concentrations that occur during anti-cyclonic conditions with transport of polluted air masses from important source areas to the south (Flanders) and east (Ruhr area) of the country. To investigate whether the origin differs as function of the modeled nitrate concentration the daily average nitrate values were categorized and the origin was plotted as function of concentration in Fig. 1. Surprisingly, the national contribution is relatively constant over the range of concentrations for the Netherlands as a whole.
Only for the populated Randstad area (e.g. Rotterdam) in the west of the country the national contribution increases with increasing concentrations. There at concentrations above 10 µg m\(^{-3}\) about 30 % of the modeled nitrate is national origin. Investigating the sectoral contributions shows that off-road transport contributions decrease with increasing nitrate levels. This can be explained by the dominant impact of international shipping on the North Sea, which is highest during westerly flows and thus clean air conditions in the Netherlands. The decreasing share of off-road transport is compensated by an increase of road transport and power generation. The modeled source attribution is location dependent. A clear example is the distribution of the foreign off-road transport contribution (Fig. 7). The modeled contributions shows a distribution with maximum absolute concentrations near the coast line, trailing off away from the coast. The maximum contributions along the coast can be explained by the formation mechanism of ammonium nitrate. Above sea, NO\(_x\) emissions are effectively converted to nitric acid by the pool of oxidants present. However, ammonia is not present in large quantities and the reaction with sea salt is not very fast, limiting the aerosol nitrate formation. However, when the air mass reaches the main land with intense ammonia emissions the formation ammonium nitrate becomes effective, explaining the maximum along the coast line.

The examples above show that system is able to track the origin of components with complex formation pathways. The system presently tracks the source contributions to oxidized (e.g. nitrogen oxides) and reduced (e.g. ammonia) nitrogen species separately. Hence, the origin of nitrate is not connected to that of ammonia, which is needed to form ammonium nitrate. Hence, ammonium nitrate is one of the few examples where the interaction between the different labeled classes is important. The present approach is chosen as the implementation is clear and results are easily explainable. An alternative could be to perform a post processing by attributing the mass of ammonium nitrate equally to the origin of ammonia and fine mode nitrate. As ammonia derives for more than 90 % from agriculture, the impact on the analysis presented above would be that agriculture would contribute almost half of the nitrate mass, whereas the con-
tributions of all other sectors are almost halved. Also, the national contribution would increase as ammonia has a shorter lifetime than oxidized nitrogen.

6 Conclusions

We have developed and demonstrated a new module to investigate source contributions to modeled air pollutant distributions. We feel that the labeling module provides more accurate information about the source contributions than using a brute force approach with scenario runs as the chemical regime remains unchanged. As many components are not inert or nowadays considered to be non-inert (e.g. primary organic matter) the ability to make a proper source allocation is enhanced. With the labeling module it is possible to do so for most species and a reasonable number of sources in one simulation. In principle, the selection of sources is flexible and can be set by the modeler. An important advantage of the new module is the reduction computation costs associated with the calculations. Not only the computer time is reduced, also the preparation of a labeling simulation requires less work than setting up N scenarios. Moreover, using a single emission database and model simulations reduces the amount of errors and recalculations. The new technology opened new research directions for e.g. the interpretation of monitoring and remote sensing data (Hendriks et al., 2012) as well as the derivation of source receptor matrices.

Acknowledgements. This study was partly funded by the 7th Framework Programme of the European Commission EnerGEO (see http://www.energeo-project.eu) and by the Dutch cooperation of model and data centers (NMDC).

References

Bobbink, R., Hicks, K., Galloway, J., Spranger, T., Alkemade, R., Ashmore, M., Bustamnte, M., Cinderby, S., Davidson, E., Dentemer, F., Emmett, B., Erisman, J.-W., Fenn, M., Gilliam, F.,


Tsyro, S. G.: To what extent can aerosol water explain the discrepancy between model calculated and gravimetric PM$_{10}$ and PM$_{2.5}$?, Atmos. Chem. Phys., 5, 515–532, doi:10.5194/acp-5-515-2005, 2005.


Fig. 1. Scatterplot of the ratio between fine elemental carbon (ec$_f$) and total primary pm in the fine mode (ppm$_f$) against the labeled fraction of ec$_f$ in ppm$_f$. This scatterplot is valid for the linear processes: Adjust layer heights, vertical diffusion, sedimentation, emission and deposition.
Fig. 2. Upper left: Concentration elemental carbon in fine mode (ec$_f$) resulting from a model simulation with only Dutch emissions. Upper right: Concentration ec$_f$ from a full model simulation corresponding to the label on Dutch emissions. Lower: The difference between those two simulations. These simulations are done with the linear processes and the non-linear advection process for the complete year 2006.
**Fig. 3.** Upper left: Concentration carbon monoxide from a full model simulation corresponding to 5% of the Dutch emissions. Upper right: Carbon monoxide concentrations from a full simulation minus concentration from a simulation with 5% reduced Dutch emissions. Lower: The difference between those two. These simulations are valid for the month July of 2006.
Fig. 4. Upper: Concentration methylglyoxal (mgly) from a full model simulation corresponding with the 5% labelled Dutch emissions. Lower left: mgly concentrations from a full simulation minus concentration from a simulation with 5% reduced Dutch anthropogenic emissions. Lower right: mgly concentrations from a full simulation minus concentration from a simulation with 5% reduced Dutch anthropogenic NMVOC emissions. These simulations are valid for the month July of 2006.
**Fig. 5.** Relative reduction of concentration NO$_2$ (left) and NO (right) by a reduction of 5% of the Dutch emissions. These simulations are valid for the month July of 2006.
Fig. 6. Relative contribution of the 5% labelled Dutch emissions to the total concentration of NO$_2$ (left) and NO (right). These simulations are valid for the month July of 2006.
**Fig. 7.** Time efficiency of the labeling module with respect to the scenario approach. On the x-axis the number of labels (sources of interest) defined. On the y-axis the relative fraction of computation time with respect to the total computation time of all needed scenario run (for each source of interest a separate scenario run).
Fig. 8. Contributions of Dutch, foreign, natural and boundary sources to different PM concentrations on five locations comparing to the monitoring stations of the BOP-campaign and one for an average over the Netherlands. Upper left: Elemental carbon aerosol, upper right: nitrate aerosol, lower: total PM10 concentration.
Fig. 9. Modeled source contributions of foreign non-road transport and Dutch road transport to nitrate concentrations (µg m$^{-3}$) across the Netherlands.