Modeling study of the impact of SO_2 volcanic passive emissions on the tropospheric sulfur budget

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Abstract. Well constraining volcanic emissions inventories in chemistry-transport models is necessary to study the impacts induced by these sources on the tropospheric sulfur composition, as well as on sulfur species concentrations and depositions at the surface. In this paper, the changes induced by the update of the volcanic sulfur emissions inventory are studied using the global chemistry-transport model MOCAGE (MOdèle de Chimie Atmosphérique à Grande Échelle). Unlike the previous inventory [Andres and Kasgnoc (1998)], the updated one [Carn et al. (2016, 2017)] uses more accurate information and includes contributions from both passive degassing and eruptive emissions. Eruptions are provided as daily total amounts of sulfur dioxide (SO_2) emitted by volcanoes in the [Carn et al. (2016, 2017)] inventories, and degassing emissions are provided as annual averages with the related mean annual uncertainties of those emissions by volcano. Information on plumes altitudes is also available and has been used in the model. We choose to analyse the year 2013, for which only a negligible amount of eruptive volcanic SO_2 emission is reported, allowing us to focus the study on the impact of passive degassing emissions on the tropospheric sulfur budget. An evaluation against OMI SO_2 total column and MODIS AOD observations shows the improvements of the model results with the updated inventory. Because the global volcanic SO_2 flux changes from 13 Tg yr^{-1} in [Andres and Kasgnoc (1998)] to 23.6 Tg yr^{-1} in [Carn et al. (2016, 2017)], significant differences appears in the global sulfur budget, mainly in the free troposphere and in the tropics. Even though volcanic SO_2 emissions represent 15 % of the total annual sulfur emissions, the volcanic contribution to the tropospheric sulfate aerosol burden is 27 %, which is due to the higher altitude of emissions from volcanoes. Moreover, a sensitivity study on passive degassing emissions, using the annual uncertainties of emissions per volcano, also confirmed the non-linear link between tropospheric sulfur species content with respect to volcanic SO_2 emissions. This study highlights the need for accurate estimates of volcanic sources in chemistry-transport models in order to properly simulate tropospheric sulfur species.

1 Introduction

Sulfur emissions come mainly from human activities (fossil fuel combustion) and volcanic activity [Andreae (1985)]. Among them, sulfur dioxide (SO_2) is a pollutant species, known to affect both human health and the environment. Because of their link to the formation of acid rain and sulfate aerosols which can induce climate forcing [Chestnut (1995), Robock (2000, 2007),

Smith et al. (2001), Schmidt et al. (2012), Kremser et al. (2016)], SO_2 emissions became a major concern in environmental policies. In some regions of the world, these policies led to strong reductions in anthropogenic SO_2 emissions in recent decades [Fioletov et al. (2016), Krotkov et al. (2016), Aas et al. (2019)]. Over North America and Europe, emissions strongly decreased between 2005 and 2015. In the East Asian region, the decrease only happened after 2010 [Sun et al. (2018)]. In contrast, over India, emissions strongly increased. And over other large SO_2-emitting regions (Mexico, South Africa, Russia or Middle East), they remained stable since 2000. However, the decrease in anthropogenic SO_2 emissions over Europe and North America was sufficient to induce an overall decrease at the global scale. Moreover, Graf et al. (1997) concluded that the efficiency of volcanic emissions to contribute to the tropospheric sulfate burden is greater than the efficiency of anthropogenic emissions, mostly because SO_2 lifetime increases with altitude and therefore has an impact for longer time periods and over larger areas. This means than in the regions where anthropogenic sulfur emissions have decreased, and more generally at the global scale, the relative proportion of volcanic sulfur emissions against the total sulfur emissions has increased.

In order to better understand the processes leading to variations in the sulfur species budget, the role of modelling is important. At the global scale, emission inventories (compilation of all available data on the globe) are used in models. Until recently, the most effective measurement instruments to assess volcanic emissions for building the inventories were the COrrelation SPECtrometer (COSPEC) ground-based instruments [details in Sect. 3.1, Moffat and Millan (1971), Williams-Jones et al. (2008)] or one of the first satellite instruments (such as the Total Ozone Mapping Spectrometer, TOMS) [Krueger et al. (1995), Seftor et al. (1997), Torres et al. (1998a, b)] but which provides only crude measurements of SO_2 column. Andres and Kasgnoc (1998) used these instruments to create one of the first global inventories of volcanic sulfur emissions. Furthermore, being compiled for the Global Emissions Inventory Activity (GEIA), it is the most widely used global dataset. For example, it has been implemented in several climate and chemistry-transport models [Chin et al. (2000), Liu et al. (2005), Shaffrey et al. (2009), Emmons et al. (2010), Lamarque et al. (2012), Savage et al. (2013), Walters et al. (2014), Michou et al. (2015)] and used in various studies on climate aerosol radiative forcing, ocean dimethyl sulfide (DMS) sensitivity or tropospheric aerosol budget [Adams et al. (2001), Takemura (2012), Michou et al. (2019), Gondwe et al. (2003a, b), Gunson et al. (2006), Liu et al. (2007)]. Subsequently, other studies using similar techniques, or building on this first inventory by supplementing it with documented sets of sporadic eruptions, have provided further global inventories [Halmer et al. (2002), Diehl et al. (2012)].

But at the time these inventories were built, techniques for measuring emission fluxes were not very accurate for the determination of volcanic sources. Indeed, ground-based instruments can only be deployed at easy-to-access volcanoes (and there are few such as, $e.g.$, Masaya) and TOMS detection sensitivity was limited only to the largest eruptions. The available inventories were therefore incomplete. The study of Andres and Kasgnoc (1998), with only one average value of all 25 years data measurements collected per volcano, reflects only a climatology without time variability. However, a lot of improvements have been made recently on satellite technologies, making possible to monitor volcanic emissions more accurately. The satellite global coverage enables to detect emission fluxes even from hard-to-access volcanoes. The improved sensitivity of the measurements has also made possible to detect not only the largest eruption fluxes but also smaller ones and persistent degassing [Yang et al. (2010), Thomas et al. (2011), Carn et al. (2013), Li et al. (2013)]. Thanks to newly developed algorithms, information on injection altitudes is available [Yang et al. (2009, 2010, 2013), Nowlan et al. (2011), Rix et al. (2012), Clarisse et al. (2014)],

reducing the uncertainties of the characterization of volcanic sources. Ge et al. (2016) highlighted the improvements made on the sulfate direct radiative forcing using both eruptive and passive degassing data in a chemistry-transport model and stressed the importance of considering the SO_2 injection altitude in volcanic emission inventories.

Carn et al. (2016, 2017) sought to compile all those new higher quality data, compared to Andres and Kasgnoc (1998), in order to provide a more representative inventory of volcanic SO_2 emissions. It is a compilation of both eruptions and passive degassing at the global scale, providing data up to a daily frequency for eruptive emissions, and a yearly frequency along with the annual uncertainty for passive emissions.

These new global volcanic sulfur inventories open the possibility of new, more detailed and accurate studies of the impact of volcanic emissions at the global scale; a stark improvement compared with studies of the last decades widely focused on major volcanic eruptions [Robock (2000)]. At the global scale, numerous studies aim at the assessment of the dispersion of sulfate aerosols and the subsequent radiative forcing [Graf et al. (1997, 1998), Gasso (2008), Ge et al. (2016)]. Regarding their impact on tropospheric composition, including air quality, several case studies at the regional scale have been analysed [e.g., Colette et al. (2010), Schmidt et al. (2015), Boichu et al. (2016, 2019), Sellitto et al. (2017)] but very few studies at the global scale have been conducted [Chin and Jacob (1996), Sheng et al. (2015), Feinberg et al. (2019)].

In this context, the objective of this work focuses on the study at the global scale of the impact of volcanic sulfur emission on the tropospheric composition, the surface concentration and the deposition of sulfur species. We aim to assess and to analyse the contribution of volcanoes to the global sulfur budget using a Chemistry-Transport Model (CTM). Here, we use the MOCAGE CTM, developed at the Centre National de Recherches Météorologiques (CNRM) [Josse et al. (2004), Guth (2015)]. Firstly, we will evaluate the changes induced by the update of the volcanic sulfur emission inventory into MOCAGE; namely from the inventory of Andres and Kasgnoc (1998) to the one of Carn et al. (2016, 2017). Secondly, the focus will be on the analysis of the volcanic SO_2 and sulfate aerosol tropospheric distribution and contribution, at the global scale, as well as the sulfur species concentration and deposition at the surface.

In Section 2, we present the configuration of simulations with the MOCAGE CTM. The new volcanic SO_2 emission inventory and its upgrades compared to the Andres and Kasgnoc (1998) one are described in Sect. 3. In Sect. 4 the setup of the simulations and the observations used to evaluate them are presented. The evaluation of the updated inventory is presented in Sect. 5. In Sect. 6, the comparison of the tropospheric and surface species concentrations between the simulations is analysed. Next, the new sulfur species distribution and budget in the atmosphere are analysed in Sect. 7. A sensitivity analysis on the passive emission sources based on the annual uncertainties provided in the Carn's inventory is carried out in Sect. 8. Finally in Sect. 9, a conclusion is given.

2 Description of MOCAGE model

2.1 General features

MOCAGE (Modèle de Chimie Atmosphérique à Grande Échelle) is an off-line global and regional three-dimensional chemistry-transport model developed at CNRM [Josse et al. (2004), Guth (2015)]. It is used for various scientific topics: impact of climate

change on atmospheric composition [e.g. Teyssèdre et al. (2007), Lacressonnière et al. (2014, 2016, 2017), Lamarque et al. (2013)], chemical exchanges between the stratosphere and the troposphere using data assimilation [e.g. El Amraoui et al. (2010), Barré et al. (2012)], operational production of air quality forecasts for France (Prev'Air program [Rouil et al. (2009)])

95 and for Europe (as one of the nine models contributing to the regional ensemble forecasting system of the Copernicus Atmosphere Monitoring Service (CAMS) European project [Marécal et al. (2015); https://atmosphere.copernicus.eu/]).

A special feature of the model makes possible to include a natural or anthropogenic accidental source, such as volcanic eruptions or nuclear explosions, during a simulation. This feature is used as part of the Toulouse VAAC (Volcanic Ash Advisory Center) of Météo-France, which is responsible for monitoring volcanic eruptions over a large area (including part of Europe

100 and Africa). In order to input an accidental emission, it is required to input the time and place (latitude/longitude), the bottom and top plume heights, the total quantity emitted as well as the duration of the emission.

2.2 Model geometry and inputs

The CTM MOCAGE can be used with global or regional resolutions based on its grid nesting capability. Each outer domain forces the inner domain at its edges (boundary conditions). The global domain has a typical resolution of $1°$ longitude x

105 $1°$ latitude (around $110\,\mathrm{km}$ x $110\,\mathrm{km}$ at the equator and $110\,\mathrm{km}$ x $80\,\mathrm{km}$ at mid-latitudes), while the regional domain resolutions are typically $0.2°$ longitude x $0.2°$ latitude (around $22\,\mathrm{km}$ x $16\,\mathrm{km}$ at mid-latitudes) and $0.1°$ longitude x $0.1°$ latitude resolution (around $11\,\mathrm{km}$ x $8\,\mathrm{km}$ at mid-latitudes).

The vertical grid has 47 levels from the surface to 5 hPa (about 35 km), with 7 levels in the planetary boundary layer, 20 in the free troposphere and 20 in the stratosphere. The vertical coordinates are expressed in σ-pressure; meaning that the model

110 levels follow closely the topography in the low atmosphere and the pressure levels in the upper atmosphere.

Being an off-line model, MOCAGE gets its meteorological fields (wind speed and direction, temperature, humidity, pressure, rain, snow and clouds) from an independent numerical weather prediction model. In practice, they can come from two meteorological models at the global scale: the IFS model (Integrated Forecast System) operated at the ECMWF (European Center for Medium-Range Weather Forecast System, http://www.ecmwf.int) or from ARPEGE model (Action de Recherche

115 Petite Echelle Grande Echelle) operated at Météo-France [Courtier et al. (1991)].

2.3 Emissions

At the global scale, anthropogenic emissions from the MACCity inventory are used [Lamarque et al. (2010)], while biogenic emissions for gaseous species are from the MEGAN-MACC inventory, also representative of the year 2010 [Sindelarova et al. (2014)]. Note that the difference between 2010 and 2013 emissions is negligible for the purpose of this study, SO_2 emissions

120 being only about 1% higher in 2010 than in 2013. Nitrogen oxides from lightning are based on Price et al. (1997) and configured dynamically according to the meteorological forcing. Organic and black carbon are taken into account following MACCity [Lamarque et al. (2010)]. DMS oceanic emissions are a monthly climatology [$1°$horizontal data Kettle et al. (1999)]. Finally, the daily biomass burning emissions available for each day in 2013 come from the GFAS daily products [Kaiser et al. (2012)]. Volcanic emissions are discussed in detail in Sect. 3.

125 In MOCAGE, with the exception of the species emitted from biomass burning [Cussac et al. (2020)], lightning NO_x [Price et al. (1997)] and aircraft [Lamarque et al. (2010)], all of the chemical species sources are injected in the first five levels of the model (up to approximately 500 m). This configuration is necessary for the numerical stability in the lowest model levels. The injection profile implemented follows an exponential decrease from the surface level of the model (including model orography): $\delta_L = 0.5\delta_{L-1}$, with δ_L the injection fraction of the mass emitted at the level L of the model. It means that the

130 majority of pollutants are emitted at the surface level and then quickly decrease with altitude. Hereafter, we will refer to "the model surface" when this configuration is used.

2.4 Chemistry and aerosols

2.4.1 Gaseous species

The MOCAGE chemical scheme is named RACMOBUS. It merges two chemical schemes representing the tropospheric and

135 stratospheric chemistry. The first one, the Regional Atmospheric Chemistry Mechanism (RACM) [Stockwell et al. (1997)], completed with the sulfur cycle [details in Guth et al. (2016)], represents tropospheric species and reactions. The second one, REactive Processes Ruling the Ozone BUdget in the Stratosphere (REPROBUS), provides the additional chemistry reactions and species relevant for the stratosphere, in particular long-lived ozone depleting substances [Lefèvre et al. (1994)].

A total of 112 gaseous compounds, 379 thermal gaseous reactions and 57 photolysis rates are represented in MOCAGE. The

140 calculation of the reaction rates is performed during the simulation every 15 min. The photolysis reaction rates are interpolated on the same 15-minute time step from a look-up table from the TUV model [Madronich (1987)]. The TUV model calculates photo-dissociation rates for both the troposphere and stratosphere. A modulation at each grid point and for all time iterations is applied as a function of the ozone column, solar zenith angle, cloud cover and surface albedo.

2.4.2 Aerosols

145 Both primary and secondary aerosols are represented in the model [Martet et al. (2009), Sič et al. (2015), Guth et al. (2016), Descheemaecker et al. (2019)]. All types of aerosols use the same set of six sectional size bins, ranging from 2 nm to 50 μm (with size bins limits of 2, 10, and 100 nm, and 1, 2.5, 10 and 50 μm).

Primary aerosols are composed of four species: black carbon, primary organic carbon, sea salt and desert dust. The first two species (black and organic carbon) depend on emission inventories while sea salts and desert dusts are dynamically emitted

150 using the meteorological forcing at the resolution of each domain [Sič et al. (2015)].

Secondary inorganic aerosols (SIA) are implemented in MOCAGE [Guth et al. (2016)]: sulfate, nitrate and ammonium aerosols. The thermodynamic equilibrium model ISORROPIA (precisely the lastest version ISORROPIA II) [Nenes et al. (1998), Fountoukis and Nenes (2007)] is used to calculate SIA concentrations in MOCAGE depending on the partition of compound concentrations, the gaseous and aerosol phases, and the ambient conditions (temperature, pressure).

155 Secondary organic aerosols are treated in MOCAGE similarly to primary aerosols with its emissions scaled on the primary anthropogenic organic carbon emissions. The scaling factor is derived from aerosol composition measurements [Castro et

al. (1999)]. The implementation in MOCAGE was done by Descheemaecker et al. (2019) in the frame of a study on data assimilation for air quality applications.

2.5 Transport

160 The transport in the model is solved in two steps. A first one explicitly determines the large-scale transport (advection) with the wind input data provided by the forcing weather model. For this purpose, a semi-Lagrangian scheme is used [Williamson and Rasch (1989)]. The second step represents the sub-grid phenomena that cannot be solved explicitly, such as convection and turbulent scattering. The convective transport is configured upon Bechtold et al. (2001) set-up. The scheme of Louis (1979) is used to diffuse the species by turbulent mixing.

3 Volcanic sulfur emissions in the model

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Volcanic emissions are composed of several gases, with the chemical composition changing from one volcano to another, depending on the geodynamical context. Sulfur species emitted by volcanoes are mainly sulfur dioxide (SO_2), and hydrosulphuric acid (H_2S) in much lower quantity. Being by far the dominant sulfur species, only SO_2 is referenced in global inventories of volcanic emissions.

3.1 Previous volcanic sulfur inventory

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The previous inventory implemented in MOCAGE is from Andres and Kasgnoc (1998), which is a study contributing to the work of GEIA. Measurements ranged over a period of about 25 years, from the early 1970s to 1997, and covered volcanic SO_2 emissions at the global scale.

A synergy between the COSPEC surface instrument and the TOMS satellite instrument was used. The COSPEC is a correla-
175 tion spectrometer initially used in pollution measurements [Moffat and Millan (1971), Williams-Jones et al. (2008)]. However, vulcanologists have adapted it to measure the quantities of sulfur dioxide in a moving air mass (here the volcanic plume). It works by comparing the amount of solar ultraviolet (UV) radiation absorbed in the plume with a standard (one sample of the background sky and two laboratory-calibrated SO_2 concentration cells). It is most commonly used under quiet to moderate eruptive conditions. On the contrary, the space instrument TOMS [Krueger et al. (1995), Seftor et al. (1997), Torres et al.
180 (1998a)], operational between 1978 and 2005, was able to detect larger eruptions. The synergy of these two instruments is therefore complementary in the development of the inventory. Although the first instrument is better adapted to the measurement of weak flares and the second to the strongest ones, a campaign dedicated to Popocatepetl in Mexico showed the good correlation between the two instruments [Schaefer et al. (2011)].

Measurements were only carried out on sub-aerial volcanoes, *i.e.*, emitting gases directly into the atmosphere. A total of
185 69 volcanoes are listed in the inventory, divided into two categories: 49 continuously erupting volcanoes and 25 sporadically erupting volcanoes. Five volcanoes belong to both categories because they had a main activity of continuous emission and also sporadic eruptive events: Aso, Augustine, Kilauea East Rift Zona, Mayon and San Cristobal.

Since the beginning of volcanic emission measurements in the early 1970s, the global activity of continuous eruptions has shown relative stability. The fluxes provided in the inventory correspond to a temporal average of all measurements for each volcano. Only three volcanoes are not concerned by this hypothesis: Etna in Sicily, Kilauea and the Kilauea Rift Zone in Hawaii, which are known as being among the largest emitters of SO_2. For those volcanoes, fluxes provided by specific studies [personally communicated to Andres and Kasgnoc (1998)] supersede the averages.

Since sporadic eruption data in Andres and Kasgnoc (1998) are not recent, it is not possible to take them into account for the recent year chosen for the MOCAGE simulation. Therefore, only continuous eruptions are used in MOCAGE and a global time-averaged SO_2 flux of 13 Tg yr^{-1} is reported.

Since no configuration was developed in MOCAGE to inject volcanic emissions aloft until this study, they were implemented similarly as the other pollution sources. Volcanic SO_2 were thus emitted at the model surface (see Sect. 2.3). However, the surface elevation of the model (orography) is mainly below the actual elevation of the volcanoes.

3.2 New volcanic sulfur inventory

With the improvements in satellite technology, an increasing number of satellites are now able to better detect the sources of volcanic SO_2: plume heights, quantities emitted and location. The most recent instruments with respect to TOMS, such as the Ozone Monitoring Instrument (OMI) and the TROPOspheric Monitoring Instrument (TROPOMI) [Theys et al. (2019)], have a higher sensitivity to detect small eruptions, but also passive degassing. Global coverage gives another considerable advantage over other measurement techniques. As a reminder, COSPEC carries out measurements from the ground and cannot be deployed on hard-to-access volcanoes.

The work of Carn et al. (2016, 2017) updates and adds complementary information to the study of Andres and Kasgnoc (1998) with a new inventory. The inventory is divided into two parts corresponding to the two types of emissions detectable by satellites.

Firstly, the eruptive emissions dataset [Carn et al. (2016) data available in Carn (2019)] is a synthesis of 40 years of daily SO_2 measurements (between 31/10/1978 and 31/12/2018) derived from 7 satellite instruments: TOMS, OMI and OMPS (Ozone Mapping and Profiler Suite) in the ultraviolet (UV), TIROS Operational Vertical Sounder (TOVS), Atmospheric InfraRed Sounder (AIRS) and Infrared Atmospheric Sounding Interferometer (IASI) in the infrared (IR) and the Microwave Limb Sounder (MLS) in the microwave range. Data from 119 volcanoes and a total of 1502 events over the period are provided. For each of these eruptions, the information given includes the location of the volcano (latitude and longitude), the date, the VEI (Volcanic Explosivity Index), the estimated SO_2 mass released (in kt), but also the height of the volcano and the height of the plume (measured if possible, estimated if not). Within our study, the additional information from Carn et al. (2016) on the injection height is used (see details here after), taking into account the height of the volcano as the base of the emissions and the height of the plume as the top of the injection.

Secondly, the passive degassing dataset is the first documented volcanic sulfur dioxide emission inventory made with global satellite measurements [Carn et al. (2017)]. It was retrieved from the observations of the OMI instrument in the UV spectrum during a long-term mission between 2005 and 2015. The high sensitivity of the instrument was a technological breakthrough

that made it possible to distinguish low SO_2 sources; $\sim$30 kt/y for persistent anthropogenic sources and lower ($\sim$6 kt/y) for volcanoes which are located at higher altitudes or at lower latitudes that benefit from more satellite observations and optimal conditions (low solar zenith angle). The volcanic SO_2 sources have been identified on the basis of 3-year averages (2005-2007, 2008-2010, 2011-2014), which implies that for a source to be characterized as persistently degassing, the emission must be relatively constant on this time scale. Annual mean emissions were calculated for each 90 volcanic source identified over the 11 years of the study. We assume in the model that emission fluxes are constant throughout the year.

Several parameters can affect the retrieval of volcanic emissions; the measurement process, the calculation algorithm or from the characterization of the type of emission. Thus, annual uncertainties are given with the mean annual emissions, for each volcanoes and each year. The total uncertainty of annual sulfur dioxide fluxes are estimated at 55 % and over 67 % for sources emitting more than 100 kt/y and less than 50 kt/y, respectively. This latter information is exploited in the sensitivity analysis (see Sect. 8). Note also that depending on the instrument used, the retrieval of the plume altitude can differ. Therefore, there are uncertainties on the altitude information provided by the inventory.

Information on the altitude of volcanoes and on the plume height in the Carn et al. (2016) inventory is used to implement a configuration to inject volcanic emissions aloft, rather than at the model surface. This is an important improvement, knowing that in some areas, depending on the model resolution chosen, the model orography may differ from the actual topography and have an impact on the transport of volcanic emissions. The new implementation sets the passively degassing emissions at the model level of the volcano altitude. For eruptions, the mass of SO_2 emitted is distributed from the model level at the volcano vent to the model level of the plume top height and follows an "umbrella" profile similar to that used in other chemistry models (Freitas et al. (2011) and Stuefer et al. (2013)). During a volcanic eruption, the emitted materials (ashes and gases) are rapidly transported vertically by the convection in the plume and most of the materials is concentrated at high altitude, giving an "umbrella" profile. In practice, the plume follows an almost linear profile with increasing altitude from the volcano vent and then opens into a parabola containing 75 % of the gases in mass into the top third of the plume.

In summary (see Table 1), the updated volcanic sulfur emission inventory includes now about 160 volcanoes ($\sim$110 in the eruptive category and $\sim$90 in the passive degassing category with 40 volcanoes in common). The availability of plume heights in this inventory allows a better representation of the injection of the volcanic emission in the model.

4   Simulation set-ups and observations

4.1   Description of the simulations

Meteorological fields are driven by the ARPEGE 3-hourly forecasts. Anthropogenic and biomass burning sources emit SO_2, whereas biogenic emissions from the ocean are assumed to occur as DMS. Oceanic DMS emissions are 19.9 Tg S yr^{-1}, while anthropogenic emissions are 48.6 Tg S yr^{-1}. For 2013, biomass burning emissions from GFAS products were relatively low, only 1 Tg S yr^{-1}.

Table 1. *Summary of the main characteristics of the previous [Andres and Kasgnoc (1998)] and the updated [Carn et al. (2016, 2017)] SO₂ volcanic emission inventories.*

	Previous Volcanic Inventory	New Volcanic Inventory	
	Andres and Kasgnoc (1998)	Carn et al. (2016)	Carn et al. (2017)
Emission type	Continuous emissions	Eruption	Passive degassing
Period	1970-1997	1978-2018	2005-2018
Instruments	COSPEC & TOMS	7 satellite instruments	OMI
Frequency	Time averaged over the period	Daily total quantity per volcano	Annual mean quantity per volcano
Information on the vertical	no information	volcano altitude	volcano altitude & plume height
Number of volcanoes	43	119	91

Table 2. *Main features of the simulations.*

	Volcanic inventory	Altitude of injection
REF	Andres and Kasgnoc (1998)	at model surface
CARNALTI	Carn et al. (2016) - eruption	from volcano vent to plume top
	Carn et al. (2017) - degassing	at volcano vent
CARN	Carn et al. (2016, 2017)	at model surface
NOVOLC	N/A	N/A

Concerning volcanic sulfur emission inventories, either Andres and Kasgnoc (1998) or Carn et al. (2016, 2017) are used. The full eruption emission database is available following Carn (2019) (https://disc.gsfc.nasa.gov/datasets/MSVOLSO2L4_3/ summary).

Four different simulations (Table 2) are carried out in order to evaluate the impact induced by the update of the volcanic SO₂ inventory in MOCAGE and to analyse its contribution to the sulfur species budget in the atmosphere at the global scale. The four simulations are run at a resolution of 1° x 1°.

The first simulation, named REF, takes into account the previous volcanic inventory [from Andres and Kasgnoc (1998)] with the injection at the model surface. The second simulation, named CARNALTI, uses the updated volcanic inventory [from Carn et al. (2016, 2017)] and the new configuration to inject volcanic emissions from the volcano altitude as described in Section 3.2. By comparing REF and CARNALTI runs, we can analyse the changes brought by the updated volcanic emission inventory with respect to the previous one. These two simulations are evaluated in Section 5 and the associated global distribution of sulfur species is compared in Section 6.

In order to distinguish between the impact of the height of emission and of the quantity of SO₂ emitted, another simulation, named CARN is run and used for the analysis of the differences between REF and CARNALTI global distribution of sulfur
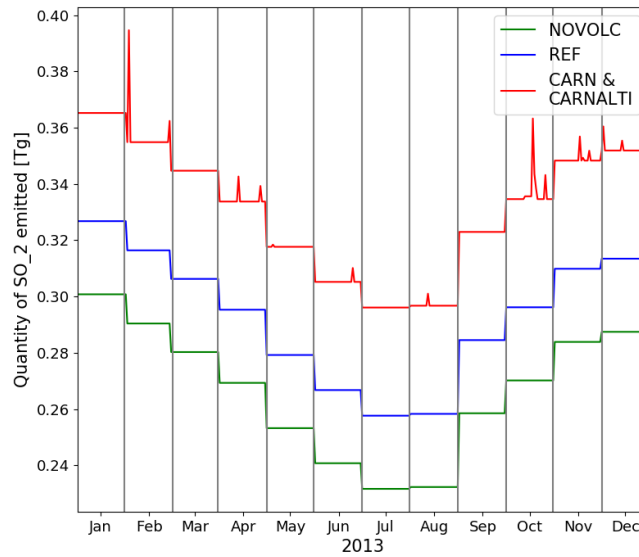
Figure 1. *Temporal evolution of 2013 SO_2 emissions in Tg; non-volcanic emissions inventory for* NOVOLC, *plus Andres and Kasgnoc (1998) volcanic emissions inventory in* REF *or Carn et al. (2016, 2017) volcanic emissions inventory in* CARN *and* CARNALTI.
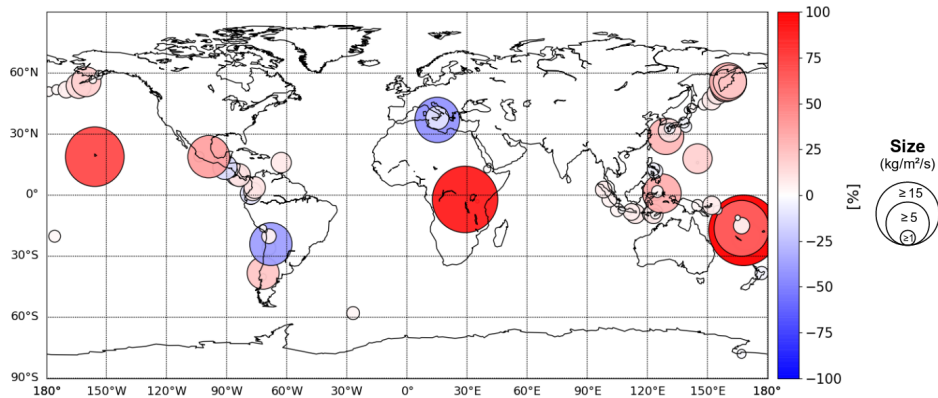
Figure 2. *2013 annual average ratio between volcanic SO_2 emissions in Carn et al. (2016, 2017) and Andres and Kasgnoc (1998) inventories. The size of the circles represents the absolute difference in $kg/m^2/s$ while the color represents the relative difference in %.*

species. Volcanic emissions are from Carn et al. (2016, 2017), like in CARNALTI but they are injected at the model surface, like in REF.

CARNALTI is run to provide a better representation of the global tropospheric sulfur. This is why it is selected for the analysis of the tropospheric sulfur budget in Section 7. In order to quantify the contribution of the volcanoes in the sulfur budget, we compare CARNALTI to the NOVOLC simulation that does not take into account volcanic emissions (only anthropogenic, biomass burning and dust).

The four simulations are run for the year 2013 with a 3 month spin-up period (from October to December 2012). In addition to being one of the years when a large amount of observational data is available globally, 2013 is chosen as the year with the lowest eruptive emission flux (Carn et al. (2016)). Figure 1 shows the volcanic emissions of the different simulations for the year 2013. We notice the monthly variation due to non-volcanic emissions (NOVOLC run in green), with less emissions during the northern hemisphere summer and the highest values in the northern hemisphere winter. Volcanic emissions from Andres and Kasgnoc (1998) are steady throughout the year, as we can see in REF run (in blue). They are lower than the volcanic emissions of CARNALTI and CARN runs (in red), with strong constant passive degassing throughout the year and a few sporadically eruptive events. Indeed, Andres and Kasgnoc (1998) SO_2 emissions are 13 Tg (or 6.5 Tg S), while the total 2013 annual emissions in Carn et al. (2016, 2017) are 23.7 Tg of SO_2 (or 11.8 Tg S), with 23.5 Tg of passive degassing SO_2 and 0.2 Tg of eruptive emission (< 1 % of the total amount of volcanic SO_2 emission, almost negligible).

Figure 2 represents spatially the difference between the previous and the new inventories. The red dots mostly show new volcanoes in Carn et al. (2016, 2017), not accounted for by Andres and Kasgnoc (1998). However, we also notice blue dots, meaning that in the new inventory, the estimated emission fluxes are reduced. Given the low number of eruptive emissions in 2013, the annual average of volcanic emissions in Fig. 2 essentially represents passive emissions.

4.2 Observations used for the evaluation of the simulations

We use for the model evaluation satellite-based instruments since they provide a global sampling. The target chemical species that we evaluate are SO_2 and aerosols, since SO_2 is the precursor of sulfate aerosols. Concerning SO_2, observations in the infrared are not suitable since passive degassing are mostly under 5 km, at altitude where such instruments have reduced sensitivity [Carboni et al. (2012), Taylor et al. (2018)]. Therefore, observations in UV-Visible are chosen. GOME-2 Metop A instrument being at the end of its lifetime, data retrievals are not good enough and presents strong artefacts, as for GOME-2 Metop B. Therefore we choose the Ozone Monitoring Instrument (OMI) instrument, being the most widely used (*e.g.* He et al. (2012), Fioletov et al. (2013), Wang et al. (2017), Wang and Wang (2020)). Moreover, the SO_2 the tropospheric column estimated from the OMI is the finest resolution and most accurate instrument in 2013 to retrieve SO_2 total columns over passively emitted volcanoes whose altitudes are generally around 2-3 km. For aerosols, there is no satellite-derived product providing information on sulfate only. Nevertheless, satellite observations of aerosols as a whole are available. Here, we choose MODIS (Moderate-Resolution Imaging Spectroradiometer) aerosol optical depth (AODs) which provides data at the global scale. MODIS AODs are known as being a robust product used in the literature for global evaluation and aerosols assimilation in models (e.g., Li et al. (2011), Dai et al. (2014), Sič et al. (2015), Guth et al. (2016, 2018)). The model comparison with MODIS AODs provides an indirect evaluation for sulfate aerosols since AODs include sulfate aerosols.

4.2.1 OMI SO_2 total column

The Aura Ozone Monitoring Instrument (OMI) level 2 sulphur dioxide (SO_2) total column product [Li et al. (2020)] was used to validate the model simulations. This product is available since 2004. The resolution of the data is 13 km x 24 km at nadir. The

305   retrieval algorithm is a principal component analysis (PCA)-based algorithm [Li et al. (2013)]. Various physical and technical causes can reduce the quality of data. Thus a pre-processing and data filtering was applied as recommended to select only the best possible observations. Pixels with large solar zenith angles (SZAs > 65°), affected by the South Atlantic Anomaly region [Richter et al. (2006)], on the edge of the swaths or the OMI row anomaly (signal suppression at certain OMI rows; see Schenkeveld et al. (2017)), pixels with a cloud fraction greater than 30% or flagged with low-confidence data are removed.

310   There are various products available in the OMI dataset since the OMI instrument has a variable sensitivity depending on altitude and the retrieval of SO_2 requires the use of an *a priori* profile. The first product selected, named "Column_Amount_SO$_2$", is an estimate of SO_2 vertical column density (VCD) and constrained by the GEOS-5 global model *a priori* profiles. Then, three specific products with adapted *a priori* profiles are also available and selected. One named "Column_Amount_SO$_2$_PBL" is an estimate of SO_2 VCD with an *a priori* profile assuming that the essential of SO_2 is in the boundary layer (within the low-

315 est 1 km of the atmosphere). Another product named "Column_Amount_SO$_2$_TRL", is almost the same as the previous one, but assuming a lower tropospheric SO_2 profile (with a center of mass altitude at 3 km). The last product selected, named "Column_Amount_SO$_2$_TRM", corresponds to an assumed middle tropospheric SO_2 profile (with a center of mass altitude at 8 km).

4.2.2   MODIS Aerosol Optical Depth

320   We use daily level 3 MODIS data (MOD08 (Terra), MYD08 (Aqua), collection 6.1) for the year 2013. Before use, we performed additional quality control and screening (Sič et al. (2015), Guth et al. (2016)). These treatments aims at minimizing cloud contamination and avoid low confidence measurements [Zhang et al. (2005), Koren et al. (2007), Remer et al. (2008)]. Moreover, all AOD values below 0.05 are automatically filtered out because Ruiz-Arias et al. (2013) highlighted the rapid growth in the relative underestimation of AODs after this threshold which leads to a mean relative error above 50 %.

325   In MOCAGE, AODs are calculated using Mie theory with the Global Aerosol Data Set's refractive indices [Köpke et al. (1997)] and extinction efficiencies derived with the Mie scattering code for homogeneous spherical particles from Wiscombe (1980).

4.3   Statistical metrics used for evaluation

In order to evaluate the model against observation data, we use the fractional bias, the fractional gross error, the root-mean-
330 square error and the correlation coefficient, following Seigneur et al. (2000).

    The fractional bias or modified normalized mean bias (MNMB) quantifies the mean between the modelled (f) and the observed (o) elements, for N observations. It ranges between -2 and 2 and varies symmetrically with respect to under and overestimation of the model. The definition is given by:

$$MNMB = \frac{2}{N} \sum_{i=1}^{N} \frac{f_i - o_i}{f_i + o_i} \tag{1}$$

The fractional gross error (FGE) quantifies the model error. It is a positive variable ranging between 0 and 2. The definition is given by:

$$FGE = \frac{2}{N} \sum_{i=1}^{N} \left| \frac{f_i - o_i}{f_i + o_i} \right| \tag{2}$$

The root-mean-square error (RMSE) is the square root of the average of the squared difference between each model and observation value. In other words, it represents a measure of the accuracy in absolute values while FGE is relative. RMSE is a positive variable, and a value of 0 (almost never achieved in practice) would indicate a perfect fit to the data. The formula is given by:

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (f_i - o_i)^2} \tag{3}$$

The correlation coefficient (R) indicates whether the variations of the model and the observations are well matched and ranges between -1 and 1. The closer the score is to 0, the weaker the correlation is. The definition is given by:

$$R = \frac{\frac{1}{N} \sum_{i=1}^{N} (f_i - \overline{f})(o_i - \overline{o})}{\sigma_f \sigma_o} \tag{4}$$

where $\overline{f}$ and $\overline{o}$ are, respectively, the model and observations mean values, and σ_f and σ_o are the standard deviations from the modelled and observed time series.

5 Evaluation of the simulations

5.1 Evaluation strategy

For the evaluation of the simulations, OMI and MODIS dataset are mapped at the model resolution ($1°$ x $1°$). The model grid points in the simulations corresponding to the filtered observation pixels (as explained in Sect. 4.2.1 and Sect. 4.2.2) are also removed. A different validation strategy is applied depending on the instrument.

Concerning OMI SO_2 total columns, similarly to other SO_2 satellite derived products, their relative uncertainties are large where the signal is low, in particular for background conditions. This is why in the literature, the SO_2 satellite comparisons and the model evaluations focus on specific areas close to SO_2 sources [*e.g.* He et al. (2012), Fioletov et al. (2013), Wang and Wang (2020)]. Similarly to these studies, our strategy is to perform the model evaluation only in the vicinity of the volcanic sources. For each volcano, based on those referenced in Carn et al. (2016, 2017), we select 9 model grid points (representing a square of $3°$ x $3°$) with the middle point being where the volcano is located (see Figure 3). Altogether it corresponds to 633 points. The mask is applied on each daily OMI SO_2 total column measurements and then we perform an annual average for each of the 633 data points. Similarly to the above mentioned studies, the results are shown as scatter plots and the statistical metrics used are the correlation coefficient and the RMSE.
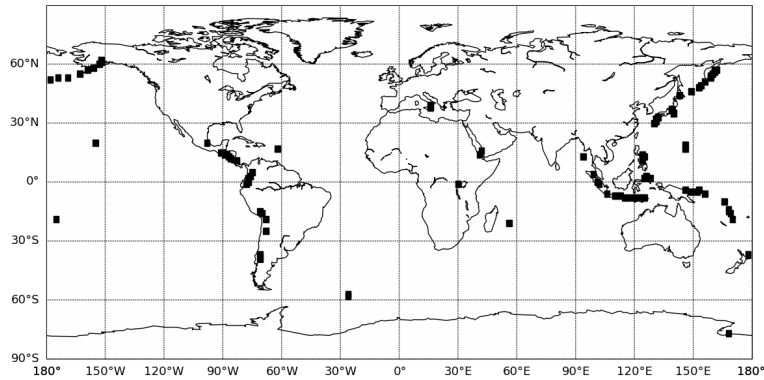
Figure 3. *Location of the selected areas where OMI SO₂ total column are selected for the validation. They correspond to 9 MOCAGE grid points around each volcano from Carn et al. (2016, 2017).*

Two methods are used in the evaluation strategy. Firstly, we choose to evaluate the model SO_2 total column against OMI "Column_Amount_SO₂" product. However, in order to test if the evaluation is sensitive to this choice, we use another approach which consists in an interpolation of OMI SO_2 observations at the altitude where the volcanic emissions are injected in MOCAGE. To do so, we use OMI products "Column_Amount_SO₂_PBL", "Column_Amount_SO₂_TRL" and "Column_Amount_SO₂_TRM", renamed hereafter "PBL", "TRL" and "TRM" respectively. Depending on the altitude of the emissions in MOCAGE, either "PBL" and "TRL", or "TRL" and "TRM", are used for the interpolation.

Concerning the AODs, a spatial validation on the whole global domain is possible against MODIS products. The evaluation at the global scale enables us to quantify the overall aerosol changes in the simulations from the use of the updated inventory with respect to the previous one. Since noticeable changes are also expected at the local scale, in the vicinity of the volcanoes, three zones are selected to complete the global scale evaluation against MODIS. These zones are chosen among the largest passive SO_2 emitters in Carn et al. (2017) and are representative of different types of changes between Andres and Kasgnoc (1998) and Carn et al. (2016, 2017) volcanic emissions inventories.

Zone 1 is centered over Central Africa and under the influence of the Nyiragongo and Nyamuragira (alt. 2950 m). In Andres and Kasgnoc (1998), this volcano is not listed. In contrast, in Carn et al. (2017), the passive degassing emission represents 2.29 Tg in 2013. No eruption is listed in Carn et al. (2016) for 2013.

Zone 2 is located in the North Pacific Ocean, around Hawaii. The volcano, based on the island, is the Kilauea (alt. 1222 m). In the REF simulation, the volcano emissions in the inventory are 0.45 Tg yr^{-1} (7 th rank of the most SO_2-emitting volcanoes in Andres and Kasgnoc (1998)). But in Carn et al. (2017), the Kilauea emissions are updated and it is the second biggest emitter with 2.17 Tg. In 2013, no eruption are recorded in Carn et al. (2016) for this area.

Zone 3 is located in the Mediterranean region, under the influence of the Etna (alt. 2711 m in the inventory) and Stromboli (alt. 870 m in the inventory). In Andres and Kasgnoc (1998), 1.48 Tg yr^{-1} are emitted by the Etna (the biggest volcanic SO_2 emitter referenced), 0.27 Tg yr^{-1} are emitted by the Stromboli and also 0.02 Tg yr^{-1} by the Vulcano. In Carn et al. (2016, 2017),

Table 3. *2013 annual statistics of* REF *and* CARNALTI *simulations against MODIS observations on specific zones.*

	Globe			Zone 1			Zone 2			Zone 3		
	MNMB	FGE	R	MNMB	FGE	R	MNMB	FGE	R	MNMB	FGE	R
REF	0.10	0.44	0.33	-0.53	0.61	0.70	0.31	0.34	0.75	0.74	0.75	0.61
CARNALTI	0.15	0.43	0.35	-0.28	0.41	0.72	0.47	0.48	0.78	0.73	0.74	0.64

only 0.65 Tg of SO_2 are emitted in 2013 in the Zone 3; corresponding to less than 0.04 Tg for Stromboli and 0.61 Tg for the Etna. The Vulcano is not in the Carn et al. (2016, 2017) inventories. In 2013, small eruptions have occurred at Etna, totalling a little less than 0.06 Tg. Therefore, in the updated Carn et al. (2016, 2017), volcanic emissions in Zone 3 are weaker than in Andres and Kasgnoc (1998).

For the evaluation of the simulations against MODIS, the statistical metrics used are the MNMB, FGE and correlation coefficient. MNMB and FGE being dimensionless, they are meaningful in all geographical regions regardless of the magnitude of the aerosol column.

5.2 Validation against OMI SO$_2$ total column

Figure 4 (top row) presents the scatter plots of MOCAGE SO_2 columns in DU from the REF and CARNALTI simulations against OMI observations based on GOES-5 *a priori* profiles. Each of the points represents an average over the 2013 year. Fig. 4 shows that the previous version of the model (REF) was not good. The coefficient correlation is low (0.12). The bias is high with a mean SO_2 measured by OMI of 0.28 DU, and of 0.08 in REF simulation. With the new volcanic inventory in CARNALTI simulation, the mean SO_2 concentration is still below OMI retrievals (0.17 DU) but much closer. We can also clearly see an improvement of the model performances with a reduced RMSE and a correlation increased up to 0.68.

To evaluate the impact of the choice of OMI product, we also show in Fig 4 (bottom row) the scatter plot when applying the interpolation at the MOCAGE altitude where volcanic emissions are injected. This method provides higher OMI estimates and therefore increases the bias with MOCAGE simulations but also improves the correlation. The conclusion is that the CARNALTI simulation provides by far better statistical results (bias, RMSE and correlation) than REF. The negative bias of MOCAGE CARNALTI with respect to OMI could be due to errors in the plume transport in the model linked to uncertainties in the meteorological inputs, to the limited number of model vertical levels, to the model chemistry/aerosol scheme but also to the uncertainties in the SO_2 emission estimates from OMI in Carn et al. (2016, 2017) and in the OMI retrieval products used for the model evaluation.

5.2.1 Validation against MODIS AOD at 550 nm

As a second evaluation step, we compare the simulations' AOD with the AOD from MODIS. Figure 5 presents for REF and CARNALTI experiments, the 2013 annual MNMB with respect to MODIS AOD observations. We can see that the equatorial
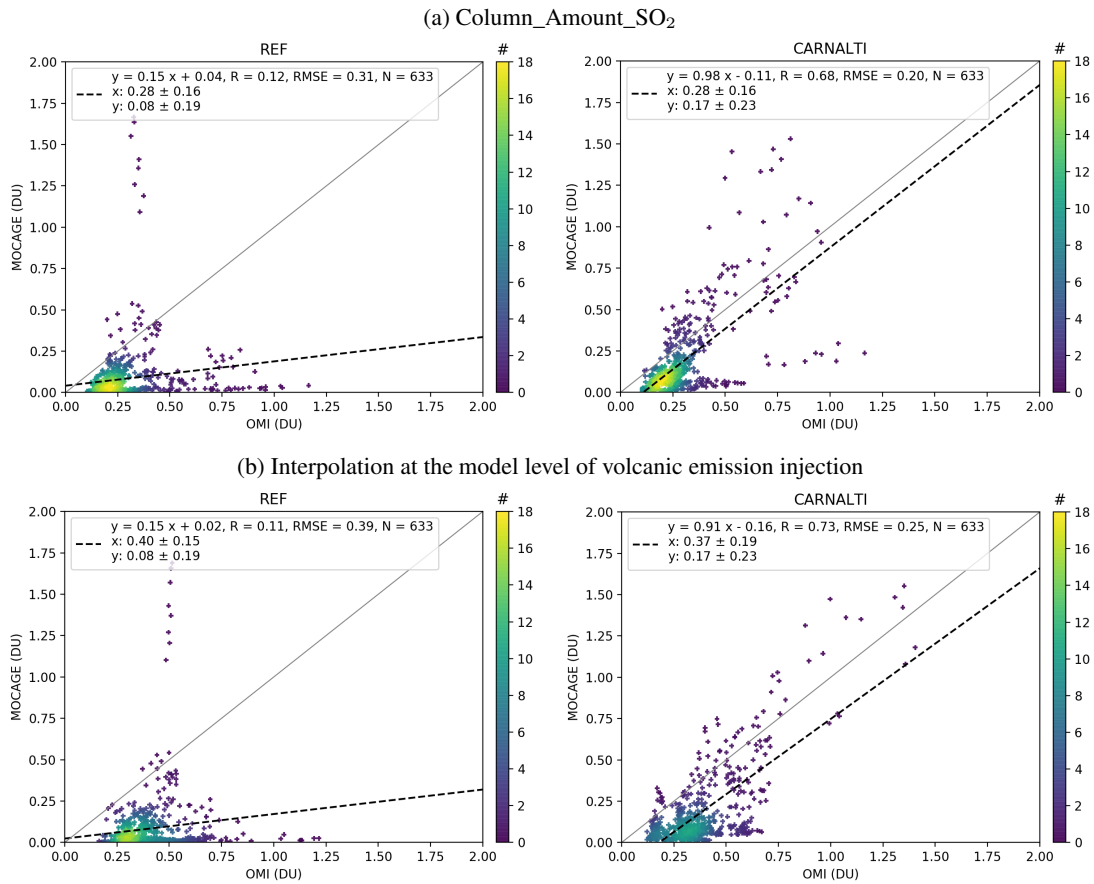
Figure 4. *Scatter plots of annual mean OMI SO$_2$ versus MOCAGE simulations (left:* REF, *right:* CARNALTI*) (a) considering total columns and (b) interpolating at the model level where volcanic emissions are injected. Also shown on the scatter plot are 1:1 line (solid grey), linear regression line (black dash), linear regression formula, correlation coefficient (R), root mean squared error (RMSE), number of collocated pairs (N), OMI mean and standard deviation in DU (x), MOCAGE mean and standard deviation in DU (y), and density of collocated pairs (colorbar).*

belt has a negative MNMB, between -0.2 and -1.2 in REF simulation, but in CARNALTI simulation, it is closer to 0; *e.g.*, in the vicinity of volcanoes in Indonesia or in central Africa. This shows an improvement of the MOCAGE AOD modeling at the global scale by updating the volcanic emissions inventory. Despite the improvement in MNMB in the areas near volcanoes, the overall score is not improved (see Table 3). Indeed, the MNMB of the Northern Hemisphere is mainly positive and almost unchanged with the new inventory [Carn et al. (2016, 2017)] where only a few volcanoes are reported. Even this small number of volcanoes, locally, leads to an increase in the already positive MNMB. Thus, globally, the average MNMB is higher in CARNALTI than in REF.

Concerning the fractional gross error (FGE), changes are also located in the vicinity of volcanoes (see Fig. S6). In those areas, especially in Central Africa and in Indonesia, the FGE is reduced from a maximum of 1.2 in REF to a maximum of 0.6
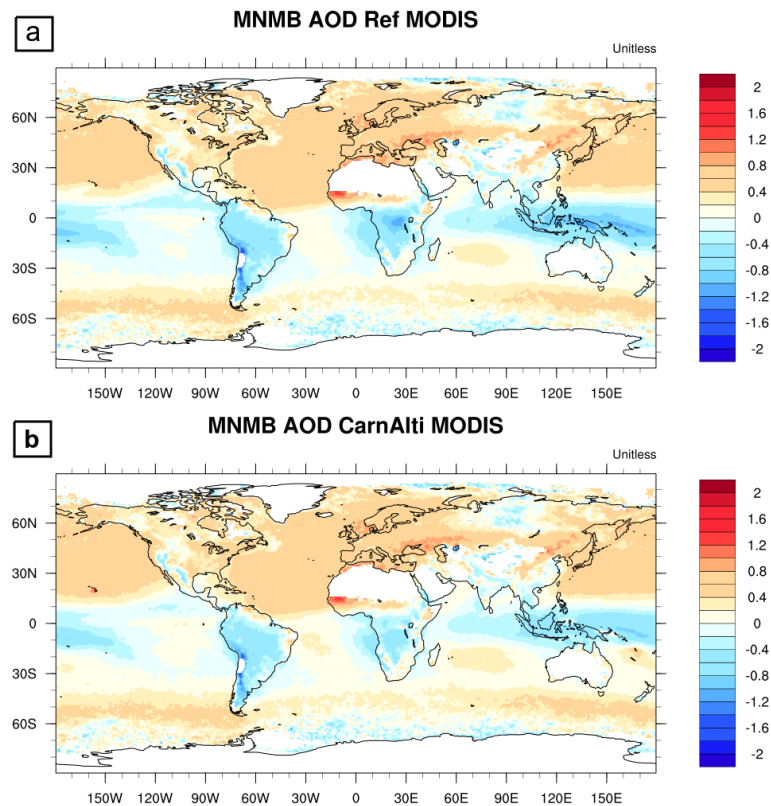
Figure 5. *Maps of the 2013 annual MNMB of aerosol optical depth against MODIS monthly observations for (a)* REF *and (b)* CARNALTI *simulations.*

in CARNALTI. Globally, the FGE score is slightly improved; 0.44 for REF and 0.43 in CARNALTI. Even if, locally in the
420   Northen Hemisphere (*e.g.* in Hawaï), the FGE score can be deteriorated in the simulation with Carn et al. (2016, 2017), at the
global scale, with the new inventory is better.

The correlation coefficient R score is better in the Northern Hemisphere (see Fig. S6). Therefore, by adding new volcano
point sources, and mostly in the Southern Hemisphere, the scores are higher in CARNALTI. The lifetime of aerosols increases
when located in higher altitude. Aerosols are better represented in the CARNALTI simulation thanks to the use of a better
425   injection altitude of SO_2 (precursor of sulfate aerosols contributing to the AOD).

By using Carn et al. (2017), the model results are improved in Zone 1. The MNMB raises from -0.53 with REF simulation
to -0.28 in CARNALTI run. Similarly, the FGE and correlation coefficient are improved. In Fig. 6 (left column for Zone 1),
the negative MNMB score in the REF simulation highlights the lack of the Nyamuragira volcanic SO_2 emissions. MNMB is
largely reduced in CARNALTI simulation.
430   In Zone 2, unlike the previous area, the MNMB is already positive. Thus, by adding more SO_2 volcanic emissions, it
increases the sulfate aerosol content leading to a deterioration of the MNMB and FGE scores (Table 3). The correlation
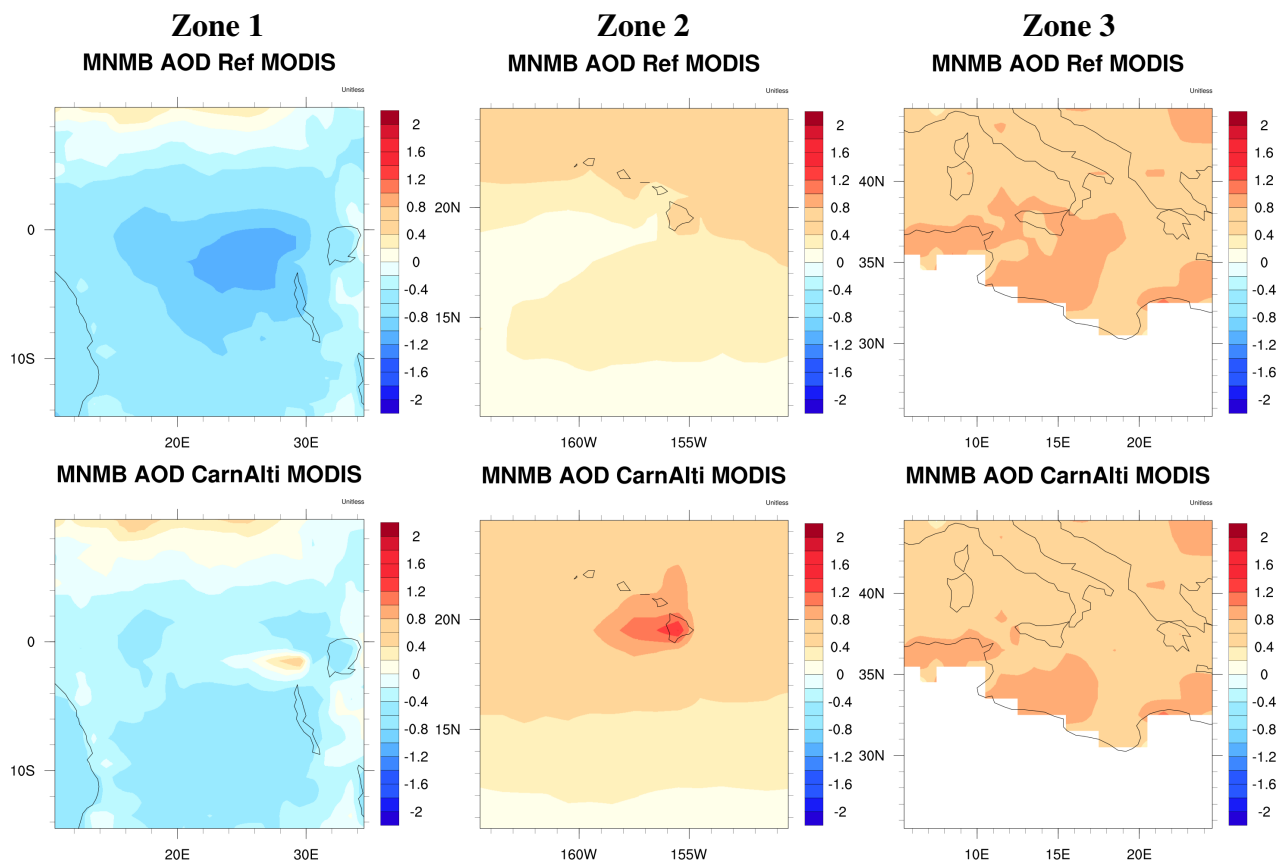
Figure 6. *Maps of the 2013 annual MNMB of* REF *and* CARNALTI *simulations against MODIS observations at the specific zones.*

coefficient increases due to a more accurate altitude where the emissions are injected in the CARNALTI simulation. Figure 6 in the middle column confirms these results. However, with the volcano being located at an altitude of 1222 m, where the sensitivity of infra-red, mostly, but also ultra-violet instruments is reduced, the estimation in the inventory for this volcano may be over-estimated.

In Zone 3, the statistical scores are almost similar for the two simulations. Indeed, in this region there are various other aerosols sources (industries, transport, dust,…) and sulfate from volcanic emissions do not dominate. Still, we can see in Fig. 6 a small improvement of MNMB between REF and CARNALTI simulations. The FGE and correlation scores are also a bit better in CARNALTI. Thus, using Carn et al. (2016, 2017) and injecting volcanic emissions at the actual altitude of the volcanoes slightly enhances MOCAGE performances.

5.3 Summary of the evaluation

The evaluation of MOCAGE performances against OMI SO_2 total column and MODIS AOD, shows an improvement of CARNALTI simulation compared to REF. The previous inventory [Andres and Kasgnoc (1998)] lacks some volcanic sources

Table 4. *Global and local (Zones 1, 2 and 3) 2013 annual mean concentrations in* REF, CARN *and* CARNALTI *simulations. Gases are in mol and aerosols in kg.*

		Mean Tropospheric Column			Mean Surface Concentration		
		SO_2	Sulfate	$PM_{2.5}$	SO_2	Sulfate	$PM_{2.5}$
		(mol m^{-2})	(kg m^{-2})	(kg m^{-2})	(mol m^{-3})	(kg m^{-3})	(kg m^{-3})
Global	REF	7.51e-6	3.25e-6	5.75e-5	7.18e-9	5.65e-10	1.26e-8
	CARN	7.92e-6	3.52e-6	5.78e-5	8.21e-9	5.91e-10	1.27e-8
	CARNALTI	8.27e-6	4.01e-6	5.85e-5	7.23e-9	5.88e-10	1.27e-8
Zone 1	REF	7.84e-6	3.93e-6	5.44e-5	4.86e-9	5.37e-10	6.03e-9
	CARN	1.28e-5	7.06e-6	5.88e-5	9.06e-9	7.51e-10	6.35e-9
	CARNALTI	1.28e-5	1.12e-5	6.51e-5	4.84e-9	1.01e-9	6.85e-9
Zone 2	REF	4.13e-6	4.43e-6	7,10e-5	4.42e-9	3.75e-10	2.27e-8
	CARN	1.45e-5	8.03e-6	7.37e-5	1.77e-8	1.11e-9	2.33e-8
	CARNALTI	2.25e-5	1.31e-5	7.92e-5	2.74e-9	1.21e-9	2.34e-8
Zone 3	REF	4.74e-5	7.66e-6	2.443e-4	4.39e-8	1.67e-9	4.75e-8
	CARN	3.22e-5	6.63e-6	2.419e-4	3.03e-8	1.25e-9	4.83e-8
	CARNALTI	3.13e-5	6.84e-6	2.424e-4	2.30e-8	1.07e-9	4.81e-8

which leads to a global underestimation of sulfur dioxide concentrations and aerosol concentrations in the tropics (e.g. in Zone 1). With the new inventory [Carn et al. (2016, 2017)] used in the CARNALTI simulation, volcanic emissions are larger. Even if in some areas the scores are deteriorated, e.g. in Zone 2 where the model is already overestimating aerosol concentrations, the scores at the global scale and in the vicinity of most of the volcanoes are improved.

6 Impact of the volcanic emission inventory update on the species concentration

SO_2, sulfate aerosols and $PM_{2.5}$ tropospheric column and surface concentrations are summarized in Table 4. In order to dissociate the effect of the quantity of SO_2 emitted and of the injection altitude, we compare the REF and CARNALTI simulations with the CARN run. The annual mean sulfur dioxide total column, at the global scale, is 8.27e-6 mol m^{-2} in the CARNALTI simulation, 10 % higher to the 7.51e-6 mol m^{-2} in REF. Regarding aerosols species, sulfate total column is 23 % higher in the CARNALTI simulation but only by 2 % for $PM_{2.5}$, because it is only partially composed of sulfate. This increase is explained by the greater amount of SO_2 emitted in Carn et al. (2016, 2017) and by the new injection configuration. At higher altitudes, the lifetime of sulfur species is longer due to slower removal processes [Stevenson et al. (2003)]. Figure 7 illustrates this concept. It shows the relative difference of sulfate tropospheric column between CARNALTI and REF experiments. We clearly see an increase in CARNALTI concentrations in the vicinity of most volcanic point sources.
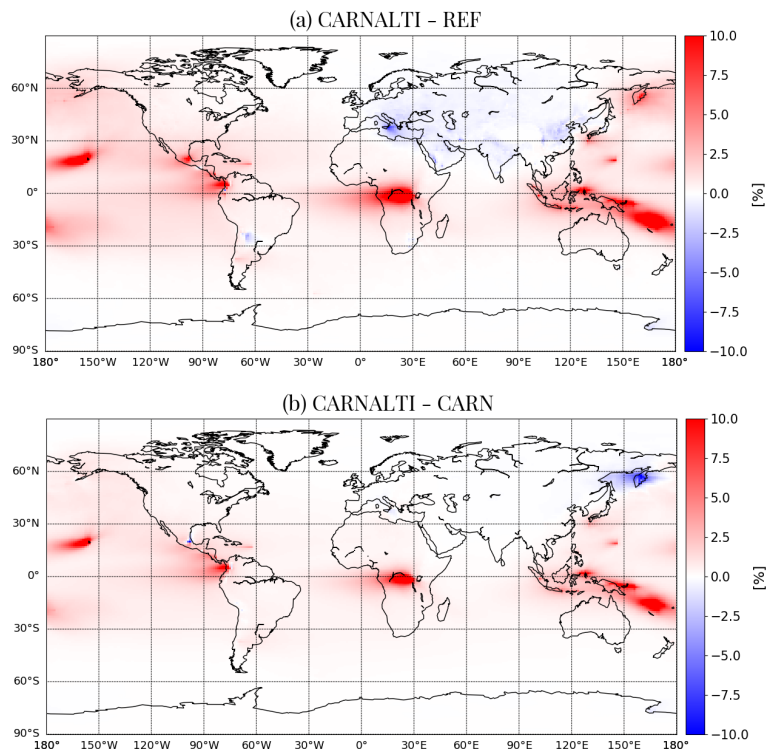
Figure 7. *2013 annual mean sulfate tropospheric column relative difference between the (a)* CARNALTI *and* REF *simulations and (b)* CARNALTI *and* CARN *simulations, in %.*

Surface concentrations, at the global scale, from the simulations show different results. The sulfur species' concentrations still increase but by less; from 5.65e-10 kg m^{-3} in REF simulation to 5.88e-10 kg m^{-3} (5 %) in CARNALTI for sulfate and from 7.18e-9 mol m^{-3} to 7.23e-9 mol m^{-3} (<1 %) for SO$_2$. In the CARN simulation results, where the volcanic emissions are injected at the model surface, we notice higher concentrations of all sulfur species at the surface. SO$_2$ and sulfate concentrations are 8.21e-9 mol m^{-3} and 5.91e-10 kg m^{-3} respectively, being around 6 % higher for each species compared with REF but 12 % higher, for SO$_2$, against the CARNALTI simulation. Thus, by injecting volcanic emission higher in altitude with the new configuration in the CARNALTI simulation, less sulfur species remain at the surface and therefore aerosols are spread further from the volcanoes (see Fig. 7b, where we can see volcanic plumes 150 to 200 km away from their source location).

By looking at the local scale, the differences between CARNALTI and REF can be very large. For example, in Zone 2, the SO$_2$ tropospheric column is 5 times larger in CARNALTI (from 4.13e-6 in REF to 2.25e-5 mol m^{-2}), 3 times larger for the aerosol sulfate total column (from 4.43e-6 to 1.31e-5 kg m^{-2}) and 3 times larger for sulfate at the surface (3.75e-10 to 1.21e-9 kg m^{-3}). In Zone 1, changes are also more important compared to the global scale, with 63 % more concentration of SO$_2$ and 185 % higher concentration of sulfate in the atmosphere and 88 % more sulfate at the surface. In Zone 3, there is less impact because it is a more polluted area.

The difference between CARN and CARNALTI SO_2 and aerosol sulfate troposheric columns are not as important as between REF and CARNALTI. Sulfur species concentrations are highest in CARNALTI, with the exception of SO_2 in Zone 3. In this highly polluted area, anthropogenic emissions are dominant. The volcanic SO_2 emitted is then more likely to compete with
475   SO_2 from other sources leading to an increase of its lifetime. At the surface, as expected, the SO_2 concentration is much higher in all zones in CARN simulation compared to CARNALTI (e.g. 85 % smaller in Zone 2 in CARNALTI compared to CARN). However, for sulfate aerosols, the surface concentrations are higher in CARNALTI run compared to CARN in Zone 1 and Zone 2. With volcanic emissions injected into the upper levels of the model, the lifetime of SO_2 increases and more sulfate aerosols are formed (as we can see in the tropospheric column) and more is found sulfate near the surface.
480   Concerning particulate matter, the impact of Carn et al. (2016, 2017) at the global scale does not present significant changes (both in the tropospheric column and at the surface), because $PM_{2.5}$ is not composed only of sulfate aerosols but are the sum of all the atmospheric aerosols with a diameter less than 2.5 μm. However, we found larger changes locally; *e.g.*, 20 % higher $PM_{2.5}$ tropospheric column concentration in CARNALTI with 6.51e-5 kg m^{-2} compared to REF with 5.44e-5 kg m^{-2}, in Zone 1. As expected, for Zone 3, all chemical species concentrations are smaller in CARNALTI compared to REF simulation, especially
485   at the surface.

7   MOCAGE sulfur budget

In this section, we calculate the MOCAGE sulfur budget and analyse the impact of the new volcanic SO_2 emissions on the tropospheric species distribution with the CARNALTI run. In order to isolate the contribution of volcanic emission from the other species concentration, we look at the difference between CARNALTI and NOVOLC simulations. The relative contribution
490   of volcanic SO_2 emissions to the species budget is defined by the quantity of species in the CARNALTI simulation subtracted from the quantity of species in the NOVOLC simulation, with respect to the total quantity of species in the CARNALTI simulation:

$$ContributionX = 100 * \frac{X_{CARNALTI} - X_{NOVOLC}}{X_{CARNALTI}} \tag{5}$$

with $X_{CARNALTI}$ and X_{NOVOLC} the annual mean concentration of the parameter X in CARNALTI and NOVOLC simu-
495   lations, respectively.

Hereafter, the parameters from NOVOLC simulation will be named "non-volcanic" parameters. On the contrary, "volcanic" parameters correspond to the parameters of CARNALTI simulation minus the quantity in the NOVOLC simulation. The CARNALTI simulation represents the total (volcanic + non-volcanic) concentration of the parameters.

7.1   Global budgets

500   The global sulfur budget simulated in CARNALTI is shown in Table 5. Annually and globally averaged SO_2 emissions, SO_2 and sulfate aerosols burdens as well as sulfur wet and dry depositions are used to calculate the sulfur budget.

Table 5. *2013 annual global mean SO$_2$ emissions, sulfur budget and deposition quantities in Tg. The contribution of sulfur species due to volcanic emissions or other emission sources are presented, in %. The efficiency is the ratio between the contribution of the sulfate burden and the contribution of the total sulfur emission attributed to a specific source. In other words, it is the fractional contribution from anthropogenic and volcanic sources to the sulfate burden.*

	Sulfur Emission	SO$_2$ Burden	Sulfate Burden	Sulfur Deposition			Efficiency
				Wet	Dry	Sedim	
Total (Tg)	81.41	0.17	0.82	42.41	27.81	9.80	-
Sources contributions to the total budget (%)							
Volcanoes	14.5	13.8	27.4	7.9	2.3	23.0	1.89
Other	85.5	86.2	72.6	92.1	97.7	77.0	0.85

Volcanic emissions are of 11.8 Tg/yr. This estimation remains in the range of previous studies which estimated volcanic emissions between 7 Tg and 14 Tg [Berresheim and Jaeschke (1983), Chin and Jacob (1996), Graf et al. (1997), Sheng et al. (2015) updated in Feinberg et al. (2019)]. However, due to lower anthropogenic emissions compared to those studies because of the recent year chosen (2013), the 15% contribution from volcanic emissions to the total sulfur emissions in CARNALTI is higher.

The global SO$_2$ burden is 0.17 Tg, slightly smaller than but closer to other studies whose values range from 0.2 Tg to 0.52 Tg [Pham et al. (1995), Chin and Jacob (1996), Feichter et al. (1996), Graf et al. (1997), Stevenson et al. (2003), Feinberg et al. (2019)]. Stevenson et al. (2003) explained that the SO$_2$ burden in a simulation can change depending on the distribution of oxidants and the deposition scheme used in the models. Moreover, those studies were done on earlier periods of time when anthropogenic SO$_2$ emissions were superior compared to 2013 in our simulation. In our simulation, 21.97 Tg S are directly removed by dry and wet deposition of sulfur dioxide, representing a percentage of almost 27 %. Thus, the transformation rate of SO$_2$ to sulfate is about 73 %; a bit higher but consistent with the studies reported above (from 50 to 66 %).

The global vertical sulfate column is 0.82 Tg S, comparable with other studies; 0.53 Tg S in Chin and Jacob (1996), 0.78 Tg S in Graf et al. (1997), 0.81 Tg S in Stevenson et al. (2003) and 0.64 Tg S in Feinberg et al. (2019).

These results confirm the non-linear contribution of the different SO$_2$ sources emissions to the sulfate burden. Indeed, volcanic sources represent almost 15 % of the total SO$_2$ emitted into the atmosphere, but they contribute 27 % to the sulfate burden. The transformation of SO$_2$ into sulfate from the other sources is not as efficient. We can note a higher efficiency for the volcanic sources, around 1.89, compared to the other sources, 0.85.

The total sulfur deposition is 80 Tg S, including 22 Tg S of SO$_2$, a little less than the total sulfur deposition in Feinberg et al. (2019), of 94 Tg S also including 22 Tg S of SO$_2$. The difference comes from the aerosol deposition, which depends on the deposition scheme and the meteorological fields which can vary depending on the considered time period. In our study, the sulfur deposition is mainly wet deposition. Precisely, the partition of each deposition fluxes are 52 % for wet deposition, 35 % for dry deposition and 12 % from sedimentation. But, sulfur deposition due to volcanic emissions is weaker than for the other sources: 8 % for wet deposition, 23 % for sulfate aerosol sedimentation and only 2 % for dry deposition. Due to the
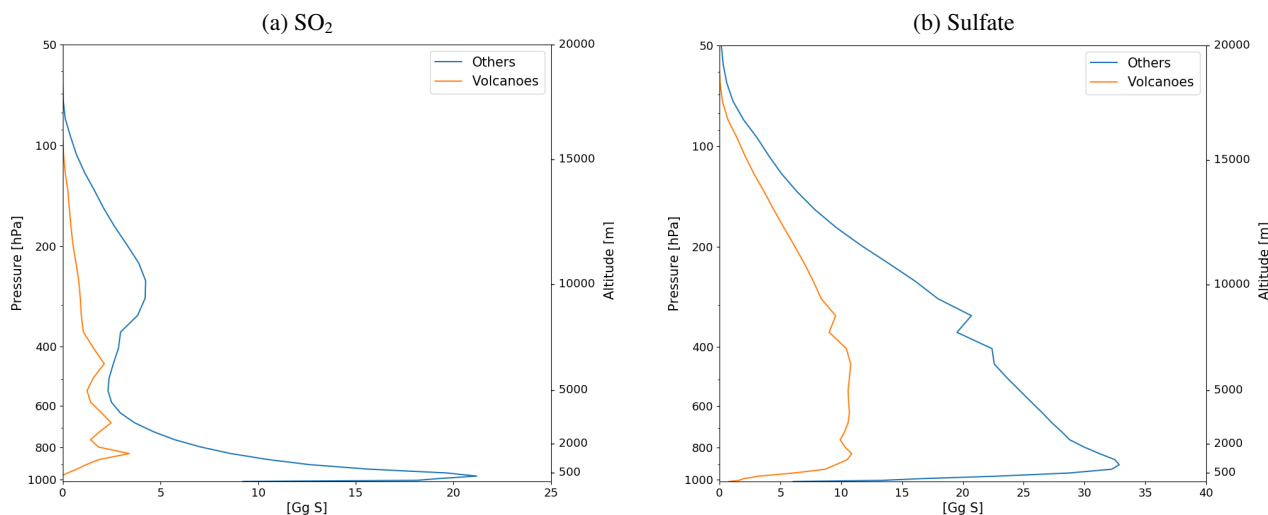
Figure 8. *2013 annual global mean vertical profile for (a) SO₂ and (b) sulfate aerosols from volcanic and other sources.*

higher altitude of injection, the atmospheric residence time for volcanic sulfur species is longer and the deposition rate lower, especially for the dry deposition.

7.2 Vertical distribution

Figure 8 shows the global and annually averaged vertical profiles for sulfur dioxide and sulfate concentrations for 2013. Anthropogenic and volcanic sources are separated to highlight the main differences between them.

Non-volcanic SO_2 dominates the entire vertical column, with a maximum at the surface linked to anthropogenic emissions, emitted at the model surface. On the contrary, the vertical distribution from volcanic SO_2 shows variations. There is no contribution below 950 hPa but there are three maxima above; one at 850 hPa (about 1500 m) due mostly to passive degassing, another around 680 hPa (about 3300 m) due to passive degassing from high-altitude volcanoes and eruptions, and the last one around 450 hPa (about 6000 m) due to high-altitude eruptions. It is noteworthy that even with few eruptive events during the year 2013, the volcanic SO_2 vertical distribution is affected by them.

Concerning sulfate aerosols, volcanic emissions are also not dominant over the entire vertical column. Non-volcanic sulfate aerosol have the highest values around 950 hPa, near the surface. For volcanic sulfate, the maximum is between 850 and 450 hPa but four times smaller than for other sources and without any specific peak associated to passive degassing or eruptive emissions. These results are different from Graf et al. (1997), which shows that the vertical distribution of volcanic sulfate aerosols is comparable to anthropogenic and biomass burning sulfate and is even dominant between 800 and 300 hPa (the altitude of volcanic emissions, mainly from eruption). This difference between our study and Graf et al. (1997) can be explained by the quantity of SO_2 emitted by eruptions. In 2013, only a few eruptive events occurred while almost 30% of volcanic emissions in Graf et al. (1997) are eruptive. Therefore, with a greater amount of volcanic emissions injected at higher altitude
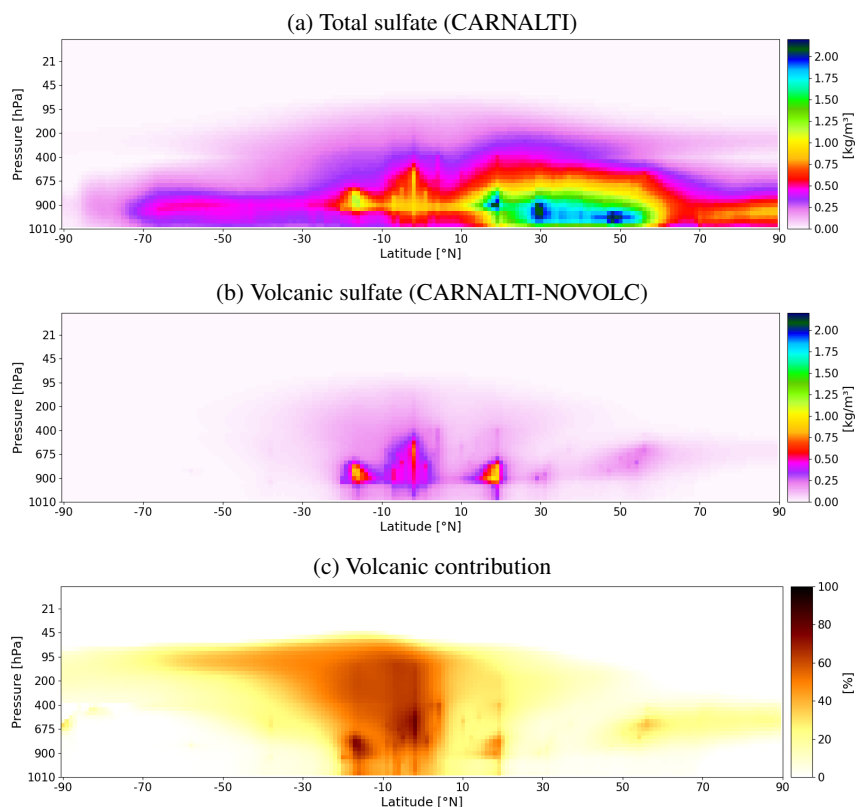
Figure 9. *2013 annual zonal mean (a) total sulfate concentration (in kg m^{-3}), (b) volcanic sulfate concentration (in kg m^{-3}) and (c) volcanic sulfate contribution (in %).*

545 in Graf et al. (1997), the potential to form sulfate aerosols is greater than in our study. This can explain the greater efficiency of 2.63 in the tropospheric sulfate burden in Graf et al. (1997) compared to 1.89 in our study.

Figure 9a represents the annual zonal mean sulfate concentration. Most of the sulfate aerosols reside in the Northern Hemisphere (between 15° N and 55° N) due to anthropogenic influence and the highest values are below 900 hPa (close to the surface). The sulfate concentrations due to volcanic emissions (Fig. 9b) are located at higher altitudes. On both sides of the

550 equator, volcanic sulfate is found between 900 and 650 hPa. Over the tropical region, the volcanoes contribution to the sulfate aerosol concentrations is larger, with a maximum of 75-80 % around 650 hPa (see Fig. 9c). We also notice that sulfate aerosols are transported by the general atmospheric circulation, up to the UTLS (Upper Troposphere - Lower Stratosphere) and from the equator to the poles, especially in the Southern Hemisphere where there are more volcanoes.

7.3 Regional distributions

555 The volcanic contribution to the global surface SO_2 concentrations is relatively low, around 14 %, but it is much higher close to the source points (see Fig. 10, top for SO_2). This is mainly due to the high altitude of emissions from volcanoes. Similarly,
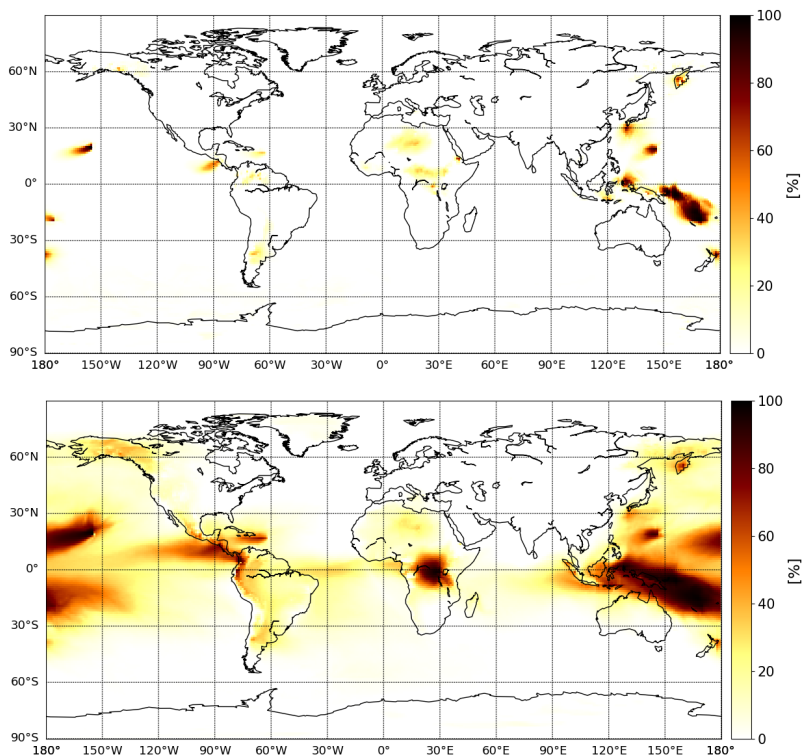
Figure 10. *2013 annual mean (top) SO$_2$ and (bottom) sulfate surface contribution due to volcanic emission (in %).*

Fig. 10 (bottom for sulfate aerosol) shows a greater influence of volcanic emissions on the sulfate aerosol concentration at the surface, almost larger than other sources in the vicinity of volcanoes. Globally, the mean contribution is of 19 %, but with a rather low, almost zero, contribution over continental areas in the Northern Hemisphere. Considering that within the
560 boundary layer, anthropogenic SO$_2$ emissions are dominant, the sulfate aerosols formed in this environment come largely from anthropogenic rather than from other sources. However, in areas with small anthropogenic sources (Indonesia, Hawaï and Central Africa), the volcanic contribution is large.

For the total column, volcanic emissions contribute much to the sulfur species burden; 24 % to SO$_2$ and 21 % to sulfate aerosols. In Fig. 11, we can see that the highest sulfate burden is located over polluted areas (East of North America, Europe,
565 Middle-East, India and China) as well as near some volcanoes and particularly over oceanic volcanoes. By looking at the volcanic contribution, we note that the sulfate aerosols due to volcanic emissions are mainly distributed over the oceanic environment in the tropics (also corresponding to volcanoes of lower altitudes). The highest contribution, 85 %, is found over Indonesia.

The annual global depositions of sulfur species due to volcanic emissions are 20 %, 5 % and 26 % for wet deposition, dry
570 deposition and sedimentation, respectively. Figure 12 represents the total sulfur deposition at the global scale and shows higher deposition fluxes over anthropogenic polluted areas, where volcanic contribution is low (see Fig. 12b). The only exception
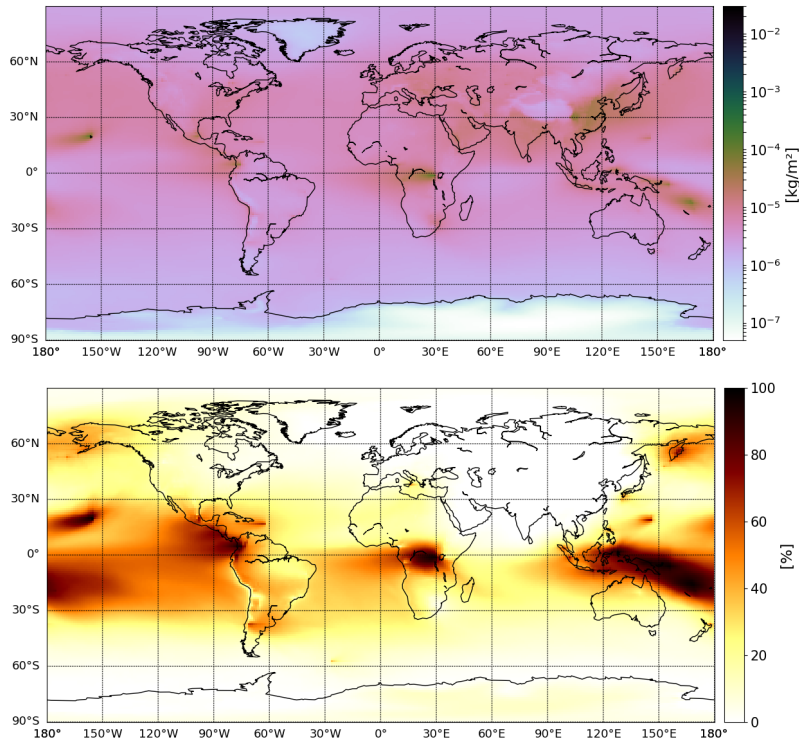
Figure 11. *(top) 2013 annual mean sulfate tropospheric column from* CARNALTI *(in kg m^{-2}) and (bottom) its contribution due to volcanic emissions (in %).*

where there are a high deposition flux and a high volcanic contribution is Indonesia. Details on the proportion of each type of deposition (wet, dry and sedimentation) are shown in Fig. S10, where we notice a weak influence of sedimentation, consistent with Table 5, compared to wet and dry depositions.

8 Sensitivity analysis on passive volcanic sources

Carn et al. (2017) provide for passive degassing not only the annual SO_2 volcanic emissions ($\overline{E_{V,Y}}$, V=volcano, Y=year), but also the associated annual emission uncertainties ($U_{V,Y}$) for each volcanic source. Thus, in this section, we aim at using this information to check the variability induced in MOCAGE sulfur budget and to analyse how it affects our conclusions from the previous section.

8.1 Description of the supplementary simulations

Three additional simulations are conducted to analyse the sensitivity of the MOCAGE model to the uncertainty of volcanic passive emissions. The first one, named CA_MIN, takes into account for each volcano the lowest estimation of SO_2 emissions. In
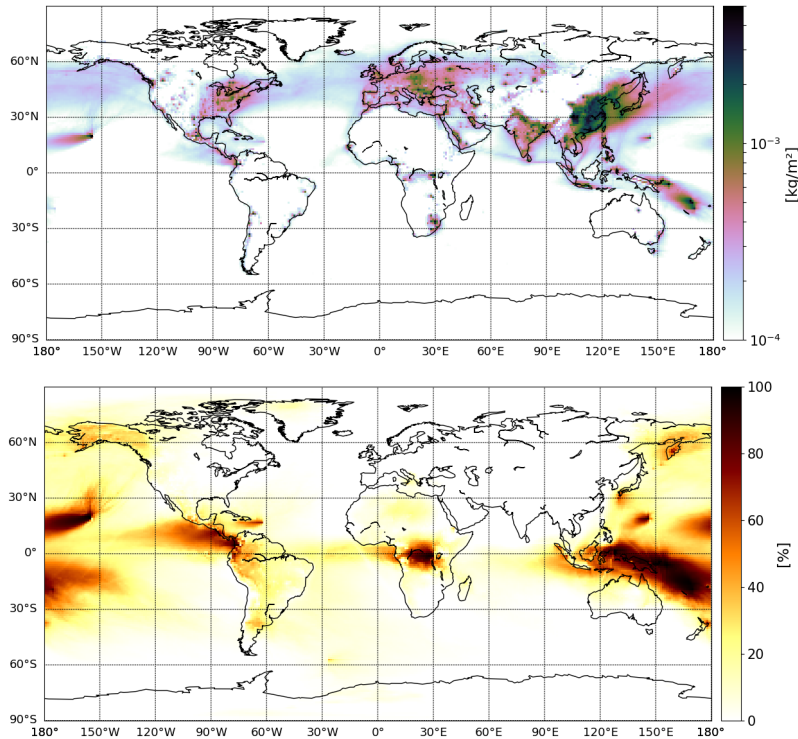
Figure 12. *(top) 2013 annual mean sulfur deposition from* CARNALTI *(in kg m^{-2}) and (bottom) its contribution due to volcanic emissions (in %).*

other words, for each volcano, we remove the annual emission uncertainty to the annual mean emission: $E_{V,Y} = \overline{E_{V,Y}} - U_{V,Y}$. In contrast, the second simulation, named CA_MAX, takes into account the highest estimation of SO$_2$ emission; we add the an-
585   nual emission uncertainty to the annual mean emission: $E_{V,Y} = \overline{E_{V,Y}} + U_{V,Y}$. Thus, both CA_MIN and CA_MAX experiments do not have daily variations due to passive degassing, but only due to eruptions. For the last one, named CA_RAND, emissions are randomly determined on a daily basis within the annual emission uncertainty interval $[\overline{E_{V,Y}} - U_{V,Y}, \overline{E_{V,Y}} + U_{V,Y}]$ following a continuous uniform distribution. Thus, daily variations are not only due to eruptions but also to passive degassing, as expected in reality. The reference simulation used, CARNALTI, is called CA from now on.

590   Figure 13 presents the 2013 temporal evolution of SO$_2$ total emission for each simulation. As in Fig. 1, we note the annual variation due to anthropogenic emissions, representing a common basis of around 70 Tg S yr^{-1} for all simulations, as well as the daily variation due to eruptions, shown by the large peaks and representing a value of 0.10 Tg S in 2013. Therefore, the differences are only due to passive degassing SO$_2$ emissions. In the CA simulation, the annual total passive degassing emission is 11.74 Tg S. In the CA_MIN, CA_MAX and CA_RAND experiments, it is 10.60, 12.95 and 11.75 Tg S, respectively.
595   Thus, there is a relative difference of 10.6 % with respect to the annual mean volcanic emissions for CA_MIN simulation but a difference of 1.4 % when considering all sulfur emissions. Similarly, volcanic emissions in CA_MAX and CA_RAND
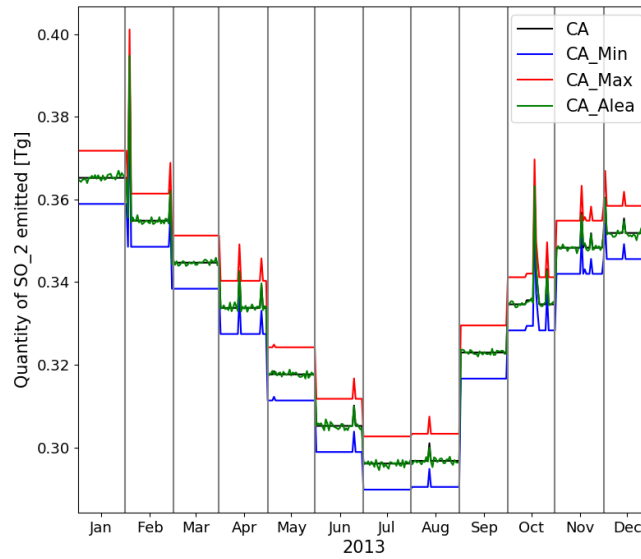
Figure 13. *Temporal evolution of 2013 SO₂ emissions, corresponding to* CA *(black),* CA_MIN *(blue),* CA_MAX *(red) and* CA_RAND *(green) simulations.*
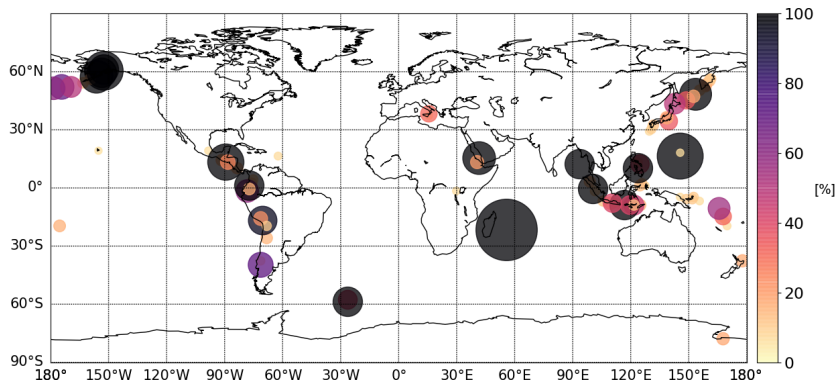
Figure 14. *Map of $U_{V,Y}/\overline{E_{V,Y}}$ ratio of SO₂ emissions (in %) in Carn et al. (2017). The size of the circles is proportional to the value of the ratio, also represented by the color.*

simulations are 9.3 % and 0.1 % higher than in CA, which represents a difference of 1.5 % and < 0.01 % respectively with respect to the total sulfur emissions.

We expect a greater sensitivity to the annual emission uncertainty at volcanoes where the proportion of the annual uncertainty with respect to the annual mean emission is close to 100 %. Figure 14 represents the percentage of uncertainty on the annual measurement of volcanic emission per volcano in Carn et al. (2017). The darker and bigger the circle is, the more important is the uncertainty compared to the mean emission.

Table 6. *As in Table 5 but for* CA_MIN, CA_MAX *and* CA_RAND *simulations.*

		Sulfur Emission	SO$_2$ Burden	Sulfate Burden	Sulfur Deposition Wet	Dry	Sedim	Efficiency
CA_MIN	Total	80.27	0.16	0.80	41.45	28.14	10.01	-
	Sources contributions to the total budget (%)							
	Volcanoes	13.33	12.68	25.30	5.74	3.43	24.59	1.90
	Other	86.67	87.32	74.70	94.26	96.57	75.41	0.86
CA_MAX	Total	82.62	0.17	0.84	42.75	28.38	10.48	-
	Sources contributions to the total budget (%)							
	Volcanoes	15.80	14.87	29.02	8.61	4.27	28.01	1.84
	Other	84.20	85.13	70.98	91.39	95.73	71.99	0.84
CA_RAND	Total	81.42	0.17	0.82	42.09	28.26	10.24	-
	Sources contributions to the total budget (%)							
	Volcanoes	14.55	13.74	27.17	7.17	3.83	26.30	1.87
	Other	85.45	86.26	72.83	92.83	96.17	73.70	0.85

8.2 Sensitivity study on the global budget in MOCAGE

As in Table 5 for CA, Table 6 presents the annual mean global sulfur budget for the CA_MIN, CA_MAX and CA_RAND simulations. Even if the total sulfur species burdens are similar in all simulations with SO$_2$ burden around 0.17 Tg S and sulfate burden between 0.80-0.84 Tg S, the contribution of the volcanic emissions to the total budget varies. In the CA experiment, the volcanic contribution to the sulfate aerosol burden is 27.40 %, but it ranges from 25.30 % in the CA_MIN experiment to 29.02 % in the CA_MAX experiment. This implies a variation in the efficiency of the model MOCAGE to produce sulfate aerosols from volcanic SO$_2$ emissions. The greatest efficiency score is 1.90 for the CA_MIN simulation, meaning that smaller amounts of SO$_2$ emitted can form sulfate more efficiently. This illustrates the non-linear relationship between the volcanic SO$_2$ emission and the sulfur budget.

Figures 15 illustrate the spatial difference in volcanic SO$_2$ contribution between CA and CA_MIN, CA_MAX and CA_RAND. The differences with CA_MIN or CA_MAX (Fig. 15a and 15b) are similar but of opposite sign. As expected, differences are located in the vicinity of volcanic point sources but especially near volcanoes with a high $U_{V,Y}/\overline{E_{V,Y}}$ ratio (see Fig. 14).

The contribution of volcanic SO$_2$ to the SO$_2$ burden is larger (less important resp.) in the CA_MAX simulation with 14.87% (the CA_MIN simulation resp. with 12.68%) than in the CA simulation with 13.80%. The difference between CA and CA_RAND are weaker. Daily variations in SO$_2$ emissions of volcanoes (CA_RAND) do not change significantly the annual mean contribution of volcanic SO$_2$ tropospheric column. The same conclusions are shown in Fig. S6 for the sulfate tropospheric column.
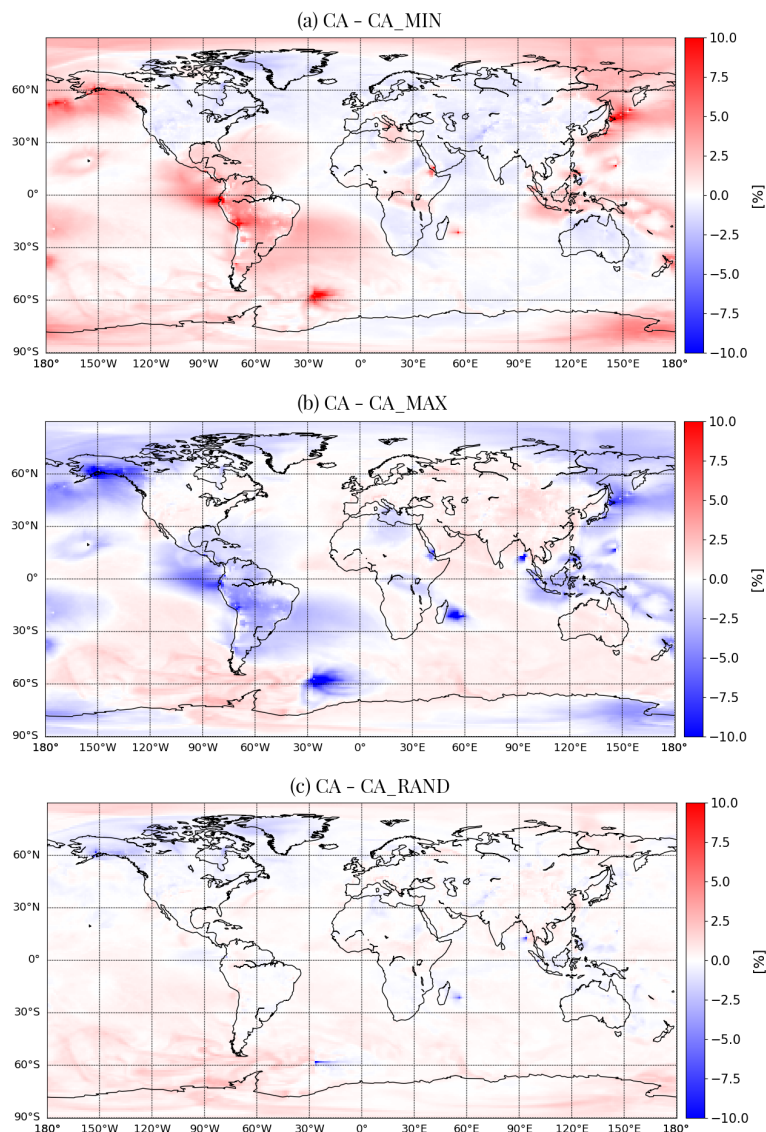
Figure 15. *2013 annual mean difference in SO$_2$ tropospheric column volcanic contribution between* CA *and (a)* CA_MIN*, (b)* CA_MAX *and (c)* CA_RAND *simulations, in %.*

620   The differences between the simulations are mostly in the deposition fluxes. Regardless of the sensitivity simulation, the sulfur deposition by dry deposition or sedimentation is higher than in the CA simulation. As an example, the sulfur sedimentation is 9.80 Tg in the CA simulation but 10.01, 10.48 and 10.24 Tg in the CA_MIN, CA_MAX and CA_RAND simulations, respectively. It represents a contribution of 22.97 % for CA and 24.59, 28.01 and 26.30 % for CA_MIN, CA_MAX and CA_RAND, respectively. Thus, sulfur deposition does not react linearly both to the quantities of volcanic SO$_2$ emitted (with respect to

625   CA_MIN and CA_MAX simulations) and to the spatio-temporal variability of these emissions (with respect to CA_RAND).

Finally, in the CA_MAX experiment, with the highest estimation of volcanic emissions, we find, as expected, a higher sulfur burden as well as higher sulfur deposition quantities. However, the CA_MIN simulation assumes the lowest estimate of volcanic SO_2 emissions and gives only a slightly lower total sulfur deposition (79.60 Tg S compared to 80.02 Tg S in CA) but with a different partition. While wet deposition is lower between CA and CA_MIN, dry deposition and sedimentation are more important in the CA_MIN simulation. Even when applying a daily variation, with nearly the same total annual quantity of volcanic SO_2 emitted (the CA_RAND simulation), we notice changes in the MOCAGE sulfur budget.

9   Conclusions

In this paper, the aim was to study the contribution of volcanic sulfur emissions on the tropospheric composition and on sulfur species surface concentration and deposition, at the global scale. Previously, the volcanic emissions inventory implemented in MOCAGE was from Andres and Kasgnoc (1998), but it has become obsolete. Therefore, a new volcanic SO_2 emission inventory, based on Carn et al. (2016, 2017), is implemented in MOCAGE. Thanks to satellite technologies, used to compile this inventory, it includes more volcanoes and gathers both eruptive emissions and passive degassing at a fine time resolution compared to previous inventories. Eruptions are provided as daily total amounts and passive degassing as annual averages with associated annual uncertainties. The inventory also provides information on the plume altitudes. A configuration to inject volcanic emission with an umbrella vertical profile was implemented in the model.

The choice was made to consider the year 2013, when quantities of volcanic SO_2 from eruptions are the lowest in the new inventory and negligible in the yearly average. Thereby, the study is focused on passive degassing emissions. Two simulations are used to assess the new version of MOCAGE using the Carn et al. (2016, 2017) emissions (CARNALTI) and associated emission heights with respect to the previous implementation based on Andres and Kasgnoc (1998) (REF).

The comparison of the MOCAGE simulations against OMI SO_2 total column and MODIS AOD shows that the statistical scores of the model were improved in the CARNALTI simulation compared to REF, especially at the local scale near the volcanoes. The global concentration of SO_2 in MOCAGE simulation is increased with the new inventory. This largely reduces the bias against OMI measurements and increases the correlation with the instrument. Compared to MODIS AOD, the underestimation in aerosol content in the tropics is also reduced. Hence, well constraining volcanic emission sources in Chemistry-Transport Models (CTM) is necessary, in order to better represent the tropospheric composition. The comparison to the MODIS AOD provides a method for validating the model results that is independent of the OMI data, which we also use for validation but was also used to help estimate the Carn et al. (2016, 2017) emissions.

We showed that considering more volcanoes (both passive degassing and eruptive types) and using a configuration to inject volcanic emissions aloft allows MOCAGE to increase sulfur species concentrations in CARNALTI compared to REF. At the surface, sulfur species concentrations and depositions were also increased, especially in the vicinity of the volcanoes, affecting air quality in these areas.

Using this new volcanic emissions inventory, we calculated the model sulfur budget in the troposphere. It shows that even if volcanic emissions represents only 15 % of the total sulfur emissions, the contribution of volcanic SO_2 emissions to the sulfur

tropospheric burden is non-linear. Indeed, volcanic sulfate burden is around 27 %, pointing out that the volcanoes efficiency
to the sulfur budget is greater than from other sources. Similarly, sulfur deposition due to volcanic emissions contributes
unequally to the total sulfur deposition, depending on the nature of deposition; e.g. volcanic sulfate aerosols sedimentation
represents the smallest proportion of the total volcanic sulfur deposition (about 12 %), but contributes significantly to the total
sulfur sedimentation from all types of SO_2 sources (about 23 %).

Moreover, the sensitivity study shows that by increasing, decreasing or including temporal variations in volcanic emission
fluxes, the global sulfur budget changes non-linearly. As an example, despite a reduction in the amount of volcanic SO_2
emitted in CA_MIN, the distribution in sulfur deposition varies, causing the decrease of wet deposition but the increase of dry
deposition and sedimentation compared to CARNALTI.

These results show that the Carn et al. (2016, 2017) inventory brings an improvement in volcanic SO_2 emissions at the global
scale. However, there are still remaining uncertainties. Even if recent important progress was made in SO_2 remote sensing, there
are various uncertainties in SO_2 retrievals from satellites of emission mass and height (for eruptions) (e.g., vertical sensitivity
of the instruments, limits of detection, assumptions in the retrieval algorithm, spatial coverage, data gaps due to clouds, . . .)
and in the methods used to derive the volcanic emissions from these retrievals. With the constant improvements of space-
borne instruments and of methods, more and more accurate volcanic SO_2 inventories will be produced in the coming years.
For example, TROPOMI instrument, with its high spatial resolution and higher-quality SO_2 dataset, could provide improved
emission inventories [Theys et al. (2019), Fioletov et al. (2020)] and could also be used to validate model in similar study as
this one, but in more recent year (2018 and later). Further gains could also be made by increasing the temporal coverage of
the satellite observations, which would enable more frequent updating of the emission inventories associated with transient
volcanic eruptions. However, this would either require more satellites to be launched into low Earth orbit or a geostationary
satellite.

In this study, we focused on one particular year. By choosing the 2013 year, we mainly study the impact of passive degassing
emissions. However, additional studies considering a year when volcanic eruptions were larger and more frequent would be
complementary; e.g. in 2014, 5.35 Tg of eruptive emissions are referenced, almost thirty times more than in 2013. It would be
interesting to compare and analyse the specific impact of eruptive emissions on the tropospheric sulfur budget. However, the
comparison of the tropospheric sulfur budget between different years cannot only be affected by the differences in volcanic
sulfur emissions. Indeed, sulfur dioxide is a soluble species and the meteorological parameters can also impact the tropo-
spheric sulfur budget; e.g., differences in precipitation can lead to changes in the wet deposition fluxes. Thus, meteorological
parameters should be taken into account to analyse the inter-annual differences.

Finally, it could also be interesting to not only compare two years of the Carn et al. (2016, 2017), but to fully study the
inter-annual variability of volcanic sulfur emissions over a longer period. Since the data are fully available over a decade
(2005-2015), this type of study would be possible.

Data availability. The new volcanic SO$_2$ inventory implemented is available for eruptive emissions on GES DISC archive (https://doi.org/10.5067/MEASURES/SO2/DATA404) and for passive degassing emissions on Carn et al. (2017) supplementary information (https://doi.org/10.1038/srep44095). Concerning data used for the validation, OMI SO$_2$ total column data can be find on the NASA database GES DISC (https://doi.org/10.5067/Aura/OMI/DATA2022). The previous volcanic SO$_2$ inventory is available upon request from the corresponding

695   author.

Supplement. The supplement related to this article is available online at: . . .

Author contributions. CL, JG and VM designed the study and the model experiments. Simulations were carried by CL with help from JG and MC. The paper was written by CL and reviewed, commented and edited by VM, JG , NT, PDH and PS. All authors approved the article.

Competing interests. The authors declare that they have no conflict of interest.

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