

Modeling natural emissions in the Community Multiscale Air Quality (CMAQ) Model–I: building an emissions data base

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Abstract. A natural emissions inventory for the continental United States and surrounding territories is needed in order to use the US Environmental Protection Agency Community Multiscale Air Quality (CMAQ) Model for simulating natural air quality. The CMAQ air modeling system (including the Sparse Matrix Operator Kernel Emissions (SMOKE) emissions processing system) currently estimates non-methane volatile organic compound (NMVOC) emissions from biogenic sources, nitrogen oxide (NO_x) emissions from soils, ammonia from animals, several types of particulate and reactive gas emissions from fires, as well as sea salt emissions. However, there are several emission categories that are not commonly treated by the standard CMAQ Model system. Most notable among these are nitrogen oxide emissions from lightning, reduced sulfur emissions from oceans, geothermal features and other continental sources, windblown dust particulate, and reactive chlorine gas emissions linked with sea salt chloride. A review of past emissions modeling work and existing global emissions data bases provides information and data necessary for preparing a more complete natural emissions data base for CMAQ applications. A model-ready natural emissions data base is developed to complement the anthropogenic emissions inventory used by the VISTAS Regional Planning Organization in its work analyzing regional haze based on the year 2002. This new data base covers a modeling domain that includes the continental United States plus large portions of Canada, Mexico and surrounding oceans. Comparing July 2002 source data reveals that natural emissions account for 16% of total gaseous sulfur (sulfur dioxide, dimethylsulfide and hydrogen sulfide), 44% of total NO_x , 80% of reactive carbonaceous gases (NMVOCs and carbon monoxide), 28%

of ammonia, 96% of total chlorine (hydrochloric acid, nitryl chloride and sea salt chloride), and 84% of fine particles (i.e., those smaller than $2.5 \mu\text{m}$ in size) released into the atmosphere. The seasonality and relative importance of the various natural emissions categories are described.

1 Introduction

Air quality modeling as practiced in the United States is often done to support air pollution management decisions that result in controls on anthropogenic pollutant sources. If the goal of this work involves secondary pollutants such as ozone or fine particles, or is regional in scope, then the air quality models that are used are typically those that include sophisticated descriptions of relevant chemical and physical processes. A major effort is required to develop emissions inputs to these models. Much of the effort involves detailed information about anthropogenic sources which are either treated individually (as so-called “point” sources) or collectively (and referred to as “area” sources; this can include “non-road” and “mobile” type sources as well). Natural emissions of biogenically produced gases or physically generated particles are also included to the extent that such emissions are considered important for the air management problem under consideration. There are several categories of natural emissions that are currently not assimilated into many regional air quality models. The reasons for this are usually associated with expectations that such emissions are not important when compared to anthropogenic emissions, or the emissions processes themselves are complex and not as well characterized as emissions that are typically modeled.

There are disadvantages to neglecting certain classes of emissions as reported by Chameides et al. (1988), Novak and Pierce (1993), DeCaria et al. (2000), Guenther et



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al. (2000), Tonnesen et al. (2006), Kaynak et al. (2008), Koo et al. (2010) and others. First, neglected emissions may, at some times and in some places, play significant roles in the photochemical evolution of the atmosphere even in the presence of anthropogenic emissions. Neglecting these emissions does not mean that they are always unimportant. Second, neglected emissions become increasingly important contributors to background levels of certain pollutants as anthropogenic emissions decrease. An air management strategy may overestimate improvements in air quality in the face of neglected emission components.

This paper describes the first part of an effort to incorporate a greater diversity of natural emissions into an air quality model that is widely used for air management studies. The effort starts by casting a wider net for emissions sources and types. The results of this work are subsequently described and the resulting natural emission components are compared with anthropogenic emissions for the same time period. Later papers will describe changes to an air quality model necessitated by the introduction of new emission sources and simulation results from that model.

2 Previous work

Much of the effort to characterize air pollutant emissions has focused on anthropogenic sources of sulfur, nitrogen oxides, and non-methane volatile organic compounds (henceforth referred to as VOC). This paper will not dwell on these sources other than to compare natural and anthropogenic emissions in Sect. 5. Note that the US Environmental Protection Agency (EPA) has an ongoing effort to characterize and quantify US man-made emissions (<http://www.epa.gov/oar/oaqps/emissions.html>). The National Acid Precipitation Assessment Program (NAPAP, 1991) conducted one of the earliest comprehensive efforts to quantify and compare anthropogenic and natural sources of emissions that influence ozone, acid deposition and haze (i.e., fine particles). That work still influences air quality modeling today.

2.1 Background

The NAPAP (1991) reported that primary natural sources of sulfur were the oceans (principally dimethylsulfide, DMS), trees (especially carbonyl sulfide, COS), soils (DMS, H₂S), crops (COS), and wetlands (H₂S, COS, and DMS). Emissions of SO₂ from volcanoes are usually sporadic and unpredictable. Emission rates for H₂S from hot springs and fumaroles (thermal vents), though small, become significant in regions where geothermal activity is high such as in Yellowstone National Park, the home to about half of the world's geysers (W. C. Evans, US Geological Survey, personal communication, 2005). The largest uncertainties – a factor of three according to NAPAP – are in soil and tree sulfur emissions. This is due to the variety of sulfur species and in-

dependent factors (i.e., meteorological, vegetation and soil type) influencing emissions, and a general lack of field data. Using data from NAPAP (1991), one can estimate that marine emissions of DMS from waters within 500 km of the US coastline average about 1400 Mg of S per annum (MgSa). Compare this with estimated annual emissions of DMS+H₂S from soils (3200 MgSa) and trees (2200 MgSa) within the continental US. Emission rates of reduced sulfur from terrestrial waters (freshwater lakes, wetlands) can also be significant but their surface coverage is very small. Finlayson-Pitts and Pitts (2000) downplay the importance of COS oxidation in the atmosphere due to its extremely slow reactivity with OH. Thus, our natural sulfur emissions modeling focused on DMS, H₂S, and SO₂.

Research by Wang et al. (1998), Guenther et al. (2000) and others has shown that the major natural sources of NO_x (NO+NO₂) include lightning, soils, and biomass burning. Ocean emissions make a considerable contribution to the global natural nitrous oxide budget. According to Denman et al. (2007), natural ocean emissions comprise 35% of the total global natural source of N₂O. Though this source is substantial it is believed to have a relatively minor impact on tropospheric chemistry (Finlayson-Pitts and Pitts, 2000). The NAPAP estimated NO_x production from lightning over North America as 1.2×10⁶ Mg of NO₂ equivalent per year with the highest emissions occurring over the southern US in summer. Guenther et al. (2000) provides an estimate of 2.9×10⁶ Mg of NO₂ equivalent per year from lightning in North America. Soil NO_x emissions depend primarily on four factors: soil moisture, vegetation cover, nutrient level and temperature. Estimates are made by scaling up from land cover type and applying measured emission fluxes adjusted for soil moisture and temperature. Continental US soil NO_x emissions have been estimated in the range of 0.4–3.6×10⁶ Mg of NO₂ equivalent per year (NAPAP, 1991). Compare this range to 1.4×10⁶ Mg of NO₂ equivalent estimated by Guenther et al. (2000) for continental US soils.

Biomass burning in the US is primarily from wildfires although agricultural fires for land clearing also occur. According to the National Interagency Fire Center (http://www.nifc.gov/fire_info/lightning_human_fires.html), the portion of wildfires caused by lightning ranged from about 11% to around 20% in the 2001–2008 timeframe with the remaining contribution considered induced by human activity. For our natural source emissions modeling purposes we focus on wildfire emissions understanding that human activity is still involved in many of these “wild” fires. Wildfires emit NO_x along with CO, NH₃, VOCs, SO₂ and fine particles. Fuel (vegetation) type, fuel moisture content, wind speed, topography, and fire type and size (area burned) all influence fuel consumption and emissions. Techniques for estimating wildfire emissions were summarized by Battye and Battye (2002) for the US EPA. From this report, the average emissions as a function of the mass of fuel burned are, by pollutant (kg emitted per 10³ kg burned): 144 (CO), 6.8 (CH₄), 3.1 (NO_x

expressed as NO₂), 6.8 (VOC carbon), 0.8 (SO₂), 0.6 (NH₃), and 12 (PM_{2.5}). The NAPAP did not estimate US emissions from fires. In recent years, detailed records of wildfires by the US Forest Service have improved the data base on wildfires and presumably increased the accuracy of fire emission estimates for air quality modeling. A key technology assisting in this wildfire inventory improvement is the SMARTFIRE (Satellite Mapping Automatic Reanalysis Tool for Fire Incident Reconciliation) system, an online system developed by a collaboration of researchers including the US Forest Service, the EPA, and Sonoma Technology, Inc. (and others). The SMARTFIRE system is designed to reconcile ground reports of fire burn information with satellite-detected fires to enhance spatiotemporal resolution and detection efficiency of fires. More information on SMARTFIRE is available at <http://www.getbluesky.org>.

Natural emissions of VOCs are almost completely associated with biogenic sources and wildfires. Vegetation emissions of various hydrocarbons have long been considered an integral component of ozone photochemistry. Sillman (1999) and others estimate that US summer biogenic emissions of hydrocarbons equal or exceed anthropogenic hydrocarbon emissions. Much biogenic VOC emissions data were collected during the 1980s and the NAPAP (1991) summarized emission rates derived from that work, including an estimate of about 27×10^6 Mg of C per year for the continental US. These VOC rates are especially important given the high reactivity of isoprene, the predominant biogenic VOC species (27% of the total in the NAPAP estimate), emitted by hardwood trees (Kinnee et al., 1997). Guenther et al. (2000) estimated about 42.8×10^6 Mg of C annually for continental US VOC with about 36% of this total attributed to isoprene.

Although wildfires are the largest natural source of primary PM_{2.5} emissions, two other sources of note are windblown dust and sea salt. Dust, usually composed primarily of weathered minerals such as carbonates, silicates and metal oxides, is produced when wind erodes exposed, dry soils. Soil particles are often lofted higher into the air through the action of convective turbulence and dust devils. Surface flux of windblown dust is controlled by soil type, vegetation cover, soil moisture and wind speed. For the purposes of this study fugitive dust produced from vehicles on unpaved roads and dust generated by agricultural tillage are considered to have anthropogenic origin and are therefore not considered a truly natural source of windblown dust. However, these components are often estimated for Regional Planning Organization (RPO) modeling inventories and the EPA National Emissions Inventory (NEI). NAPAP (1991) estimated the total US emission of alkaline soil dust (including roads) at about 19×10^6 Mg yr⁻¹. Emission estimates for these fugitive dust sources are considered to have high uncertainty.

Sea salt is generated by the action of wind blowing over the ocean, generating waves and promoting the formation of bubbles (from trapped air) and spray from breaking waves. Small droplets are produced from spray and as bubbles burst

upon reaching the surface. Evaporating droplets leave behind salt particles composed primarily of chlorides. Athanasopoulou et al. (2008) reviewed past efforts at estimating sea salt particle fluxes and applications to atmospheric modeling. They opted to use the methodology of Clarke et al. (2006) as extended to smaller particle sizes by Gong (2003) for open ocean sea salt aerosol (SSA) production. In addition, they used the methodology by De Leeuw et al. (2000) for estimating SSA flux in surf zones. Both methodologies estimate size-dependent SSA flux as a function of 10-m wind speed, correcting for the effect of relative humidity on SSA diameter. Athanasopoulou et al. (2008) input data from both flux methods to the CAMx air quality model (ENVIRON, 2004) to examine SSA transport and chemical evolution over the Greek archipelago. They concluded that their selected methodologies for estimating SSA flux produced realistic results for SSA over land, with the contribution to total PM₁₀ (60% maximum) and PM_{2.5} (10% maximum) dropping rapidly with distance from the coast.

2.2 Natural emissions in the EPA CMAQ model

One widely used air quality simulation tool, originally developed by the US EPA, is the Community Multiscale Air Quality (CMAQ) Model (Byun and Schere, 2006). Currently, the treatment of natural emissions in CMAQ includes the following: biogenic sources (i.e. vegetation and soil), windblown dust (via special treatment with modules external to the CMAQ system), wildfires, and sea salt. The Sparse Matrix Operator Kernel Emissions (SMOKE) modeling system (CMAS, 2008) is a data preprocessor that provides emissions input files for CMAQ and other air quality models. For biogenic sources SMOKE produces VOC emissions from vegetation and NO emissions from soil based on land cover-specific emissions factors and meteorology. Enhancements to windblown dust emissions modeling have recently been introduced such as the development of a windblown dust model for the Western Regional Air Partnership Regional Planning Organization (WRAP) RPO produced by ENVIRON International Corporation (Mansell et al., 2006) that allows treatment of dust emissions based on meteorology and soil/landscape cover offering improvements over the more generalized estimates typically found in emissions inventory development.

Traditionally, SMOKE treats wildfire emissions as low-level area sources. However, recent RPO inventories have been developed to provide daily fire emissions from point source wildfires along with hourly plume rise for vertical allocation of their emissions. SMOKE has also been adjusted to handle other fire-specific characteristics such as fuel loading and fire burn area (<http://www.smoke-model.org>).

Beginning with CMAQ version 4.5, a new aerosol module was included to perform calculations of sea salt emissions from open oceans. This module produces emissions as a function of wind speed and relative humidity (<http://www.epa.gov/cmaq/>).

[//www.cmaq-model.org](http://www.cmaq-model.org)). Prior to this version of CMAQ, sea salt – if included at all – was only handled separately outside the model. Such an approach required a special model to produce emission estimates.

2.3 Global emissions

An emerging modeling system suitable for both global and regional simulation of air emissions from terrestrial ecosystems is the MEGAN (Model of Emissions of Gases and Aerosols from Nature) system (<http://acd.ucar.edu/~guenther/MEGAN/MEGAN.htm>). MEGAN is designed to be a replacement to the EPA Biogenic Emissions Inventory System (BEIS) modeling system and has been used in a number of recent studies investigating VOC flux from vegetation including Guenther et al. (2006), Helmig et al. (2007), Sakulyanontvittaya et al. (2008) and others. Two examples of global atmospheric chemistry models are the GEOS-Chem (<http://acmg.seas.harvard.edu/geos/index.html>) and the GO-CART models (http://data.eol.ucar.edu/codiac_data/ace-asia/docs/model_chem.GO-CART.html). These models have been used by Park et al. (2004, 2006), Chin et al. (2002), the US RPOs and others in research that incorporated natural emission source influences. Common emissions inventories employed by these global models include the GEIA, EDGAR, and POET databases. Natural source emissions provided in the GEIA (Global Emissions Inventory Activity) data set or from GEIA-affiliated data sets include releases from soils, biomass burning, animals, oceans, volcanoes, vegetation, and lightning. Further information on the GEIA emissions inventories is available from <http://www.geiacenter.org>.

Other examples of global emissions inventories for natural sources include the POET (<http://www.aero.jussieu.fr/projet/ACCENT/POET.php>) and EDGAR (<http://www.mnp.nl/edgar/>) global data sets. Natural source emissions from POET include global emissions estimates for ozone precursor gases from biogenic and oceanic sources (Granier et al., 2005). EDGAR provides natural emission estimates for N₂O from soils, grasslands, and oceans/wetlands; non-methane VOCs from natural vegetation; CH₄ from soils, grasslands, natural vegetation, and oceans/wetlands; NH₃ from wild animals, soils, and oceans; NO_x from soils and lightning; and CO from vegetation and oceans/wetlands (Olivier et al., 2002). Uncertainty estimates for these natural emissions range from 50 to 100% (Olivier et al., 2002).

There are numerous examples of previous continental to global scale air quality modeling that have incorporated these global emissions data sources. Park et al. (2003) – focusing on background levels of organic carbon (OC) and elemental carbon (EC) particles over the United States – used an inventory of global anthropogenic OC and EC emissions developed by Cooke et al. (1999) for 1984 and modified for their 1995 and 1998 simulations. While simulating sulfate and nitrate aerosol, Park et al. (2004) utilized global fossil

fuel and industrial emissions from the GEIA data base as described by Benkovitz et al. (1996). Marine emissions of DMS were calculated from monthly gridded seawater concentrations from Kettle et al. (1999). Park et al. (2004) also used ammonia emissions from the GEIA inventory of Bouwman et al. (1997). The EDGAR inventory was used by Sudo et al. (2002) to model global air quality with the CHASER model. They also obtained ocean emissions from Müller (1992) and biomass burning data from Hao and Piu (1994).

A common feature of all these inventories is their coarse spatial resolution. This is not necessarily bad given their use in global models but it does present some challenges if they are used in higher resolution models. Ongoing efforts to examine natural emissions impacts over the continental US requires emissions primarily from North America and surrounding oceans. Outside this area model boundary conditions must be derived from outputs from global models that use coarsely-resolved global emissions and will have a minor influence on results for the continental US. The widespread acceptance and ease of access to the GEIA data base makes it highly attractive for model applications.

3 Adding natural emissions to the CMAQ Model

The CMAQ Model was selected for use in an ongoing study of the role of natural emissions in US air quality. Recently, CMAQ input emissions files for 2002 were prepared to simulate aerosols (Tesche et al., 2006). These files include detailed anthropogenic sources, biogenic sources of VOCs and NO_x, windblown dust, and wildfires (MACTEC, 2008). Methods and data bases used to add other natural emissions to those already prepared for 2002 are described below.

3.1 Lightning NO_x

Emissions of NO_x from lightning discharges have been included in a number of air quality modeling exercises, including Novak and Pierce (1993), Sudo et al. (2002), Park et al. (2004), Heald et al. (2006), Kaminski et al. (2007) and others. NO_x from lightning represents a significant portion of the annual NO_x emissions budget, with estimates of between 5 and 11% in the studies just cited. The effects of lightning NO_x can be local – from cloud-to-ground discharges that inject NO_x directly into the boundary layer – and regional due to the long-range transport of lightning NO_x discharged into the upper troposphere. This is one source of NO_x that has traditionally not been treated in regional simulations using CMAQ and its emissions processing system SMOKE.

Recently, much effort has gone into developing methods to estimate NO_x produced by lightning. One approach is to use a satellite optical detector to estimate the occurrence of lightning flashes (Nesbitt et al., 2000). From this one can estimate LNO_x by assuming a per-flash NO_x production rate. The method suffers from incomplete global coverage and

low detector sensitivity, as well as the usual problems associated with estimating the ratios of cloud-to-ground (CG) to intra-cloud (IC) flashes and the vertical distribution of LNO_x . Another method uses cloud information such as convective cloud occurrence and top height generated by a meteorological model and then assumes flash rates, vertical distributions and NO_x formation rates associated with the simulated convection (Pickering et al., 2006). The US National Aeronautics and Space Administration (NASA) is experimenting with using remote sensing of atmospheric chemical signatures to estimate lightning NO_x for use in real-time air quality forecasting (<http://nasascience.nasa.gov/earth-science/applied-sciences/resources/CMAQr1.pdf>). The method used here is very similar to that of Kaynak et al. (2008) who used lightning flash data from the National Lightning Detection Network (NLDN). Kaynak et al. (2008) horizontally allocated lightning-derived NO_x based on whether clouds were simulated to occur in the CMAQ model grid cell associated with the measured lightning flash or in adjacent cells whenever no clouds were simulated in the flash cell. The latter occurrence was very rare (<1.5% at most). Vertical allocation of emissions was based on published lightning profiles and modeled cloud top heights. Kaynak et al. (2008) made CMAQ model simulations for July–August 2004 with and without lightning NO_x and found that lightning could contribute significantly to ground level ozone in some circumstances. Another recent study that also determined that lightning could contribute significantly to ozone levels, particularly in the southeastern US, is the work performed by Koo et al. (2010). Koo et al. (2010) simulated lightning NO assuming a static emission rate of 9.3 Kg N per CG flash from Holler and Schumann (2000). Koo et al. (2010) assumed 30 million flashes per year for North America from Orville et al. (2002) then applied a multiplier of 2.8 IC flashes per CG flash from Boccippio et al. (2001) resulting in NO simulation for 114 million flashes. Koo et al. (2010) then used hourly convective precipitation, cloud depth, and pressure to make spatiotemporal allocations of these lightning emissions.

In the present work 2002 NLDN flash data were acquired for the CMAQ modeling domain (see the Appendix A). Rather than simply applying an assumed static factor for flash emissions rates as has been done elsewhere, the emissions rates for flashes in the 2002 simulation used an algorithm to estimate flash-specific emissions rates. This approach, referred to as the mean peak current method, is a variation on the methods described by Price et al. (1997), DeCaria et al. (2000), and Ott et al. (2010). The basis of this approach is CG flash data from the NLDN data set. We introduce separate treatment of positive flash return strokes and use all NLDN peak current values (rather than setting a minimum threshold) as recorded in the data set. Details of calculations used to estimate LNO_x emissions are summarized in the Appendix A. The vertical allocation of LNO_x was not tied to simulated cloud top heights but was instead fixed. This assumption seems reasonable given the unknown

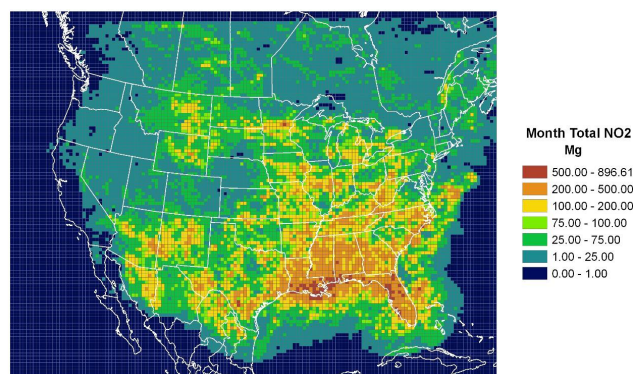


Fig. 1. Total lightning NO_x emissions (Mg of NO_2) calculated for July 2002 across the atmospheric modeling domain derived from NLDN data.

and suspected limited skill of the meteorological model to accurately estimate convective cloud top heights for 36 km grid cells. The resulting spatial distribution of LNO_x emissions at the time of the expected summer peak is shown in Fig. 1.

3.2 Reduced sulfur

Globally, reduced sulfur emissions from the oceans represent at least 20% of the total sulfur budget (Park et al., 2004). Reduced sulfur occurs primarily in the form of DMS but may also exist as H_2S , COS, dimethyldisulfide, carbon disulfide and methyl mercaptan (Finlayson-Pitts and Pitts, 2000). Over the oceans DMS emissions are an important source of marine aerosols (Kreidenweis et al., 1991). DMS and H_2S are both significant over land – especially from soils, certain vegetation and surface waters – jointly accounting for over half of total reduced sulfur emissions in the US (NAPAP, 1991). The balance of naturally emitted sulfur is mostly as COS, a relatively slow-reacting compound (Finlayson-Pitts and Pitts, 2000). Natural sulfur sources are small in magnitude compared to anthropogenic sources of SO_2 (NAPAP, 1991; Finlayson-Pitts and Pitts, 2000; Park et al., 2004) but may become important, in the absence of significant SO_2 , as a source of atmospheric sulfate.

3.2.1 Oceans

Ocean DMS emissions were taken from the annual mean flux of sulfur from the oceans as delineated by Kloster et al. (2006). Monthly emissions were produced using the ocean DMS temporal profile shown in the Appendix A. This monthly profile derives from the work of Kettle and Andreae (2000). Kloster et al. (2006) stated that their ocean DMS results were in good agreement with the work performed by Kettle and Andreae (2000). Hourly ocean DMS emissions were produced for a representative day using the SMOKE

default temporal profile for natural sources also shown in the Appendix A.

Estimated ocean H₂S flux utilized the emissions factor (see Appendix A) recommended by Andreae and Jaeschke (1992). Given the limited and varying information available on marine H₂S temporal variations, we decided to simply allocate the emissions using the factor from Andreae and Jaeschke (1992) into a single representative day with equal allocation of the emissions into each hour of the day. Further refinement of this method may be appropriate for future modeling of this source.

3.2.2 Coastal wetlands

Sulfur emission factors applied in this modeling for coastal wetlands (as shown in the Appendix A) are an average of the results from a number of studies on sulfur flux from salt marsh and *Spartina Alterniflora* including Steudler et al. (1985), Giblin and Wieder (1992), and DeLaune et al. (2002). According to the US Department of Agriculture (<http://plants.usda.gov/>), *Spartina Alterniflora* distributions are found in twenty US coastal states and along the eastern coastlines of Canada. Significant distributions of *Spartina Alterniflora* are not found along the west coast of Canada and it is not native to Mexico nor were there any area distributions for this grass readily available for Mexico. A number of land cover datasets were used to select coastal areas of land cover classes relevant to existing wetland sulfur emission factors. These land cover datasets included US NLCD (National Land Cover Database) for 1992 and 2001, GAP land cover (for Florida only), and land cover from the NOAA Coastal Services Center. More recent NLCD datasets were not available for many areas during this project. For more information and access to these datasets visit <http://www.epa.gov/mrlc/>. Emissions were produced for a single representative day using the SMOKE default natural source temporal profile (see Appendix A). Incorporating monthly/seasonal variability to these emissions should be explored for future work.

3.2.3 Freshwater

Emission factors applied here for H₂S and DMS from freshwater lakes (see Appendix A) in Canada were selected by taking values from the lower range of freshwater sulfur flux estimates derived from several studies including Richards et al. (1991), Bates et al. (1992), Giblin and Wieder (1992), Gröne (1997), Bodenbender et al. (1999), and Sharma et al. (1999). A representative day of H₂S and DMS emissions was created from these emission factors and the SMOKE default temporal profile for natural sources. This representative day was then assumed for all remaining days of the ice-free season (i.e. April–November following Sharma et al., 1999).

Sulfur emissions for the Great Lakes were estimated for a representative day using the same H₂S and DMS emissions and temporal factors that were applied for the afore-

mentioned Canadian interior freshwater lakes. However, no frozen lake periods were assumed because of the small fraction (no more than approximately 11% ice cover for all lakes combined in the Great Lakes for 2002; see <http://www.glerl.noaa.gov/data/ice/atlas/>) of total surface area covered by winter ice.

3.2.4 Other sources

Other natural sources of reduced sulfur that were modeled include the Great Salt Lake (GSL), soils, and geothermal sources such as volcanoes and fumaroles. Given the lack of available information on natural atmospheric emissions from the GSL, it was decided to apply DMS emissions factors for Canadian interior lakes similar in composition to the GSL. Comparisons were made to lakes described in Richards et al. (1994). Patience Lake located in southern Saskatchewan was the closest match in composition to the GSL (assuming seawater as the closest proxy to the GSL). Fluxes of DMS were created using the mean DMS flux for Patience Lake as given in Richards et al. (1994). The H₂S emissions factors for GSL were more problematic. Because no data were available, the same H₂S emission factor used for the Great Lakes was applied to the GSL. A freshwater H₂S factor should be very conservative when applied to the GSL. A representative day of DMS and H₂S emissions were created using the same natural source temporal allocation profile used for other species. Emissions factors used for the GSL are shown in the Appendix A.

We estimated reduced sulfur emission from soils based on the ratio of soil NO_x and reduced sulfur (DMS and H₂S) budgets estimated by NAPAP (1991). The resulting conversion factors for H₂S and DMS were then directly applied to the SMOKE v2.5 BEIS3.14 processor results for soil nitric oxide (used as a modeled spatial surrogate) to estimate reduced sulfur fluxes.

Geothermal fluxes of natural sulfur emissions including volcanoes, fumaroles and hot springs were also simulated. Geothermal sources and emissions factors simulated here are shown in the Appendix A. Selection of volcanoes for this modeling was based on the availability of source-specific emissions factors. Though Mexico contains four major volcanoes none of them are found in our modeling domain, thus, only other Mexican geothermal features were included. Likewise, in Canada only non-volcanic geothermal features were included in the modeling domain. The SMOKE default natural source temporal profile was used to produce a representative day of emissions for geothermal SO₂ and H₂S.

3.3 Ammonia

Ammonia plays two important roles in the evolution of natural air quality. Its alkalinity affects the pH balance of cloud droplets and droplet acidity influences various chemical reaction rates (National Research Council, 1983). Ammonia

gas reacts with nitric acid in the presence of water to produce ammonium nitrate aerosols (Finlayson-Pitts and Pitts, 2000). The natural sources of ammonia included in this simulation were oceans, wild animals, and wildfires. Wildfires are not discussed here as they are detailed further in Sect. 3.5. It is important to note that while native soils are considered a viable natural source of atmospheric NH_3 (about 17% of the natural NH_3 total emissions for the contiguous US in 2001 according to Park et al., 2004) soil NH_3 emissions are also difficult to quantify because soils can both emit and absorb ammonia (NAPAP, 1991). This creates ambiguity over whether soils are a net sink or source of ammonia. Therefore, soil NH_3 was not included in the current natural emissions modeling inventory.¹

The method applied here for modeling natural emissions of NH_3 from wild animals was based on the work described by Mansell (2005). The Carnegie Mellon University (CMU) Ammonia Model version 3.6 (Davidson et al., 2004) was used to estimate the annual total wild animal (i.e. bear, deer, and elk) NH_3 emissions from the contiguous US (emissions factors shown in the Appendix A). These wild animal emissions were considered temporally invariant consistent with both Davidson et al. (2004) and Mansell (2005). Annual emissions were subsequently converted to a representative day with no hourly variation. The wild animal NH_3 emissions for Canada were also estimated using the CMU factors. These factors were applied to activity data acquired (where available) from government environmental agencies for each Canadian province in the modeling domain. Additionally, Canadian populations (i.e. activity) of moose were also incorporated. The CMU emission factor for elk was used as a surrogate for moose. The same type of temporally invariant representative day of emissions as created for wild animals in the US was also created for Canadian wild animals. Given the limited information on wild animal populations within Mexico, we decided to use the GEIA wild animal NH_3 emissions data set in our emission inventory. Annual totals from the GEIA data base were handled in a similar fashion as performed in our work for US wild animal emissions.

Annual emissions estimates of natural NH_3 from oceans were taken from the GEIA database and once again the default SMOKE natural source temporal profile was used to create a temporally invariant representative day of ocean NH_3 . Further research on the temporal allocation of ocean NH_3 emissions for the refinement of default temporal profiles used in this simulation is needed.

3.4 Chlorine

Atmospheric chlorine is very reactive in its various forms. Reactive Cl_2 photolyzes into atomic Cl, which further reacts with many substances, including VOCs, DMS, and H_2S .

Many of these reactions were included in the revised CMAQ CBIV gas chemical mechanism starting with version 4.5 (<http://www.cmascenter.org>) and were based on the work of Tanaka et al. (2003a). Reaction of Cl with O_3 produces ClO which in turn reacts with HO_2 to produce HOCl (Atkinson et al., 2007). Furthermore, HOCl photolyzes to produce Cl and OH. Thus, in coastal areas where natural chlorine is most abundant it can play an important role in photochemistry (Knipping and Dabdub, 2003). In addition, hydrochloric acid is readily absorbed into cloud droplets, influencing pH and aqueous reactions.

Aside from volcanic emissions of HCl, the primary source of natural chlorine is sea salt dechlorination by N_2O_5 (Behnke et al., 1997). This process yields ClNO_2 that photolyzes to release Cl or reacts with OH in the gas phase to form HOCl (Atkinson et al., 2007). Simulating the reaction of N_2O_5 with sea salt in an air quality model requires modifications to the model's physical/chemical aerosol module in order to accurately capture the relationship between gases and particles. The current study does not attempt modifications to the CMAQ aerosol module. Instead, ClNO_2 is incorporated into CMAQ by adding estimates of equivalent near-surface ClNO_2 flux as a surrogate for complex gas-aerosol interactions.

Natural sources of chlorine emissions estimated for our 2002 inventory include sea salt aerosols (fine and coarse chloride) and gaseous HCl and ClNO_2 from the oceans. Annual totals of HCl and ClNO_2 from oceans were taken from the GEIA Reactive Chlorine Emissions Inventory (RCEI) (Erickson et al., 1999). Monthly totals were derived from factors in Erickson et al. (1999). These factors produce lower emissions in the summer months and peaks in the winter. The GEIA ClNO_2 fluxes for the ocean waters around North America were modified to reflect the fact that they include the influences of both natural and anthropogenic NO_x emissions. This was done by determining the ratio of natural to anthropogenic NO_x over land near the oceans and then partitioning the GEIA ClNO_2 fluxes proportionately into natural and anthropogenic components. This produced a much lower estimate of natural ClNO_2 flux than would otherwise have been estimated and avoided the introduction of artificially high NO_2 levels over the oceans once daytime ClNO_2 reacts with OH to liberate NO_2 . Monthly natural weighting factors calculated ranged from 12 to 17% (denoting subtle seasonal variations for the split) with the remaining 83 to 88% assigned as the monthly anthropogenic weighting factors. A representative day of emissions for each month was subsequently created with hourly allocations based on Keene et al. (2007) and Osthoff et al. (2008). Temporal profiles applied for HCl and ClNO_2 are shown in the Appendix A.

¹Ammonia emissions from fertilized agricultural soils are considered to be an anthropogenic source.

3.5 Wildfires

Wildfire emissions have been traditionally treated as stationary area sources. However, this has changed such that wildfire emissions are now treated as point sources with enhanced temporal modeling including day-specific emissions and hourly plume rise (<http://www.smoke-model.org/version2.5>). Fortunately, the intricacies of handling fire records for inclusion in air quality modeling were already performed for 2002 by the various RPOs. The appropriate fire records were identified in the RPO inventories using the source classification code (SCC) for wildfires and by assuming all other fires to be anthropogenic. Canada wildfires were selected as available from the 2002 MARAMA (Mid-Atlantic Regional Air Management Association) and Environment Canada 2000 inventories while Mexico wildfires were selected from Phase III of the 1999 Mexico National Emissions Inventory (NEI). The US, Canadian and Mexican wildfire inventories were subsequently processed with SMOKE. Area source fires were processed for representative days while point source fires were processed for each day of the episode to utilize daily meteorology (needed for plume rise calculations). This approach closely mimics the method applied for the VISTAS RPO 2002 simulations described in Morris et al. (2004). Pollutants produced from wildfires in our modeling included VOCs, CO, NH₃, NO_x, SO₂, PM_c, unspicuated PM_{2.5} and fine particulate EC, nitrate, OC, and sulfate.

3.6 Primary PM_{2.5}

Natural sources of primary PM_{2.5} include wildfires, sea salt, and windblown dust. Treatment of wildfires and sea salt were discussed previously. Windblown dust has been poorly represented, when not ignored completely, within regional air quality models such as CMAQ (Pierce et al., 2003; Choi and Fernando, 2008). An accurate treatment of windblown dust must incorporate complex atmospheric processes with varying land surfaces and soil data (Pierce et al., 2003). We acquired WRAP RPO 2002 CMAQ-ready natural source windblown dust emissions data. These were created for the WRAP RPO by ENVIRON International Corporation using its own Windblown Dust Model (Mansell et al., 2006). The model considers the necessary parameters for adequate windblown dust emissions estimation as described in Pierce et al. (2003) and Choi and Fernando (2008). The principal model drivers of emissions are wind speed (wind friction velocity) and soil characteristics with varying land cover type (Mansell et al., 2006). The model produces coarse and fine primary particulate emissions.

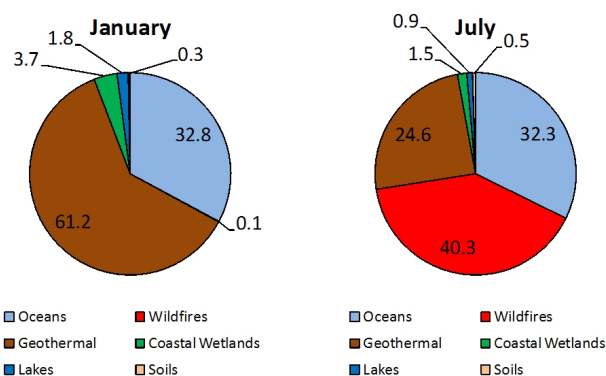


Fig. 2. Percent distribution of natural gaseous sulfur (SO₂+DMS+H₂S) emissions by source.

Table 1. Monthly emissions totals for naturally occurring sulfur gases within the atmospheric modeling domain.

Species	Jan Mg	Jul Mg
SO ₂	5.2×10^3	1.2×10^5
DMS	3.5×10^4	8.8×10^4
H ₂ S	2.2×10^4	2.1×10^4
Total sulfur	3.0×10^4	7.3×10^4

4 Natural emissions for 2002

4.1 Gaseous sulfur

Distributions of naturally-emitted gaseous sulfur species by source type are shown in Fig. 2 for January and July 2002. Geogenic and ocean sources dominate in January. Wildfires play a much greater role in July. Contributions from inland surface waters, soils and coastal wetlands comprise only about 6% of the total emissions in January and 3% in July.

Domain-wide emission totals for the three natural gaseous sulfur species are summarized in Table 1 for January and July of 2002. January natural emissions of gaseous sulfur compounds are dominated by DMS and H₂S with SO₂ emissions being nearly a factor of 10 smaller. This reverses in July with SO₂ emissions being almost 40% greater than DMS. Figure 3 illustrates the spatial distribution of total sulfur emissions for January and July. The primary difference is due to the abundance of wildfires that occur in summer. Total natural gaseous sulfur emissions in July (7.3×10^4 Mg) are about 2.5 times higher than in January. These results suggest that winter sulfur aerosol formation might be controlled by the slow oxidation of reduced sulfur compounds into SO₂ and eventually sulfate. It also implies that natural sulfate aerosol levels should be greater near the US coastlines in winter because the predominant precursors are emitted over the oceans. Wildfires should shift the higher natural sulfate levels inland in summer because of the increase in both SO₂ and sulfate aerosol emissions from fires.

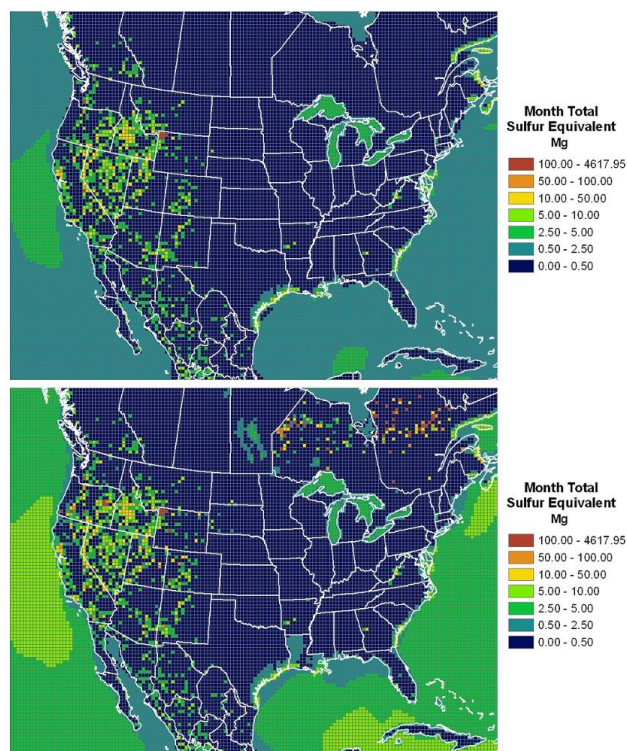


Fig. 3. Total natural gaseous sulfur (SO₂+DMS+H₂S) emissions, expressed as Mg of S, calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.

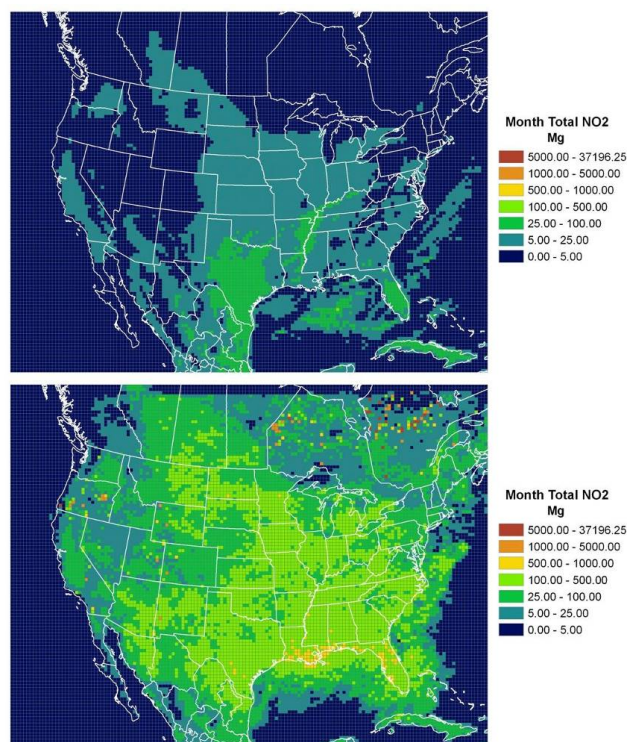


Fig. 5. Total natural NO_x emissions, expressed as Mg of NO₂, calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.

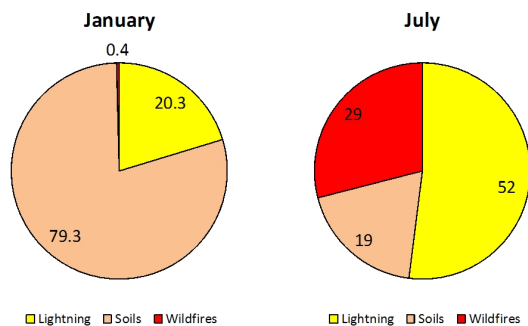


Fig. 4. Percent distribution of natural NO_x emissions by source.

4.2 NO_x

The relative amounts of natural NO_x emissions are illustrated in Fig. 4. Most January NO_x comes from soils but lightning accounts for more than half of all emissions in July. Wild-fire emissions are nearly nonexistent in January but comprise almost 30% of total NO_x in July. Figure 5 shows the January and July spatial distributions of natural NO_x emissions. Lightning NO_x is mostly confined to the south-central US and Gulf of Mexico in the middle of winter. By midsummer there are few locations that do not experience some lightning and NO_x emissions are especially dense over the southeast-

ern US. Wildfires considered naturally-caused that were inventoried for this 2002 simulation occurred primarily across the western US, in Florida and eastern Canada.

Table 2 quantifies the seasonal breakdown of natural NO_x emissions along with that of other gaseous nitrogen species. Nitrogen in the form of NO_x is far greater than that emitted as ammonia or nitryl chloride (the latter is confined to marine coastal areas). The data suggest that lightning could occasionally make significant contributions to ozone formation in summer over the southern US, consistent with that reported by Kaynak et al. (2008). Lightning NO_x also has the greatest potential to contribute to nitrogen deposition in summer over the southern and eastern US.

4.3 NH₃

Natural ammonia plays a small role in neutralizing acid aerosols and contributing to ammonium nitrate aerosol formation. Monthly emission totals listed in Table 2 indicate the July emissions are much greater than those of January. The higher summer emissions (July: 1.2×10^5 Mg = 7×10^9 moles) are such that the total moles emitted of natural NH₃ exceeds those of natural sulfur species by a factor of 3. The potential exists for natural NH₃ to neutralize natural oxidized sulfur in some locations, especially over the oceans. Also, maritime ammonia emitted in the vicinity of lightning NO_x,

Table 2. Monthly emissions totals for naturally occurring nitrogen gases within the atmospheric modeling domain.

Species	Jan Mg	Jul Mg
NO _x ^a	9.2×10 ⁴	1.4×10 ⁶
NH ₃	2.6×10 ⁴	1.2×10 ⁵
ClNO ₂	8.6×10 ²	2.2×10 ¹
Total nitrogen	5.0×10 ⁴	5.2×10 ⁵

^a Expressed as NO₂.

and wildfire ammonia and NO_x co-emissions, have the potential to form ammonium nitrate aerosols in select locations.

4.4 Gaseous chlorine

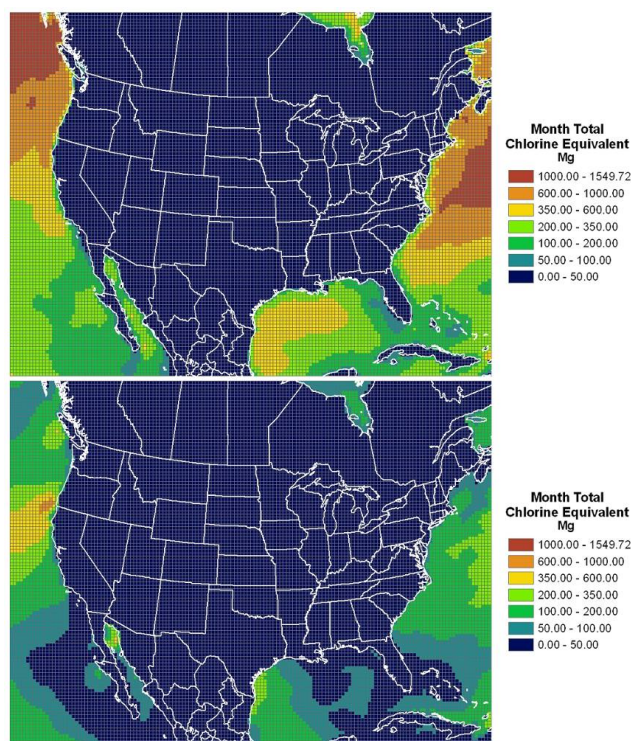
Natural chlorine emissions are such that airborne gaseous chlorine compounds are confined to air over the oceans and in coastal zones (Fig. 6). Table 3 indicates that hydrochloric acid is the predominant gas species with nitryl chloride a distant second. Both originate from sea salt which is more abundant when winds are strong. This means that winter and spring are generally the periods when sea salt and associated chlorine gas species will have the largest contribution to air quality and atmospheric chemistry. Also, ClNO₂ levels will be much higher when natural and anthropogenic emissions are combined because this analysis tied ClNO₂ to nearby NO_x levels. When total gas+aerosol chlorine is considered sea salt is dominant accounting for 99% of all airborne chlorine.

4.5 VOCs and CO

Volatile organic compounds are dominated by biogenic sources but significant contributions are also made by wildfires. The US Southeast shows the highest VOC emissions throughout the year (Fig. 7) but wildfires increase western US emissions in summer. July VOC emissions for 2002 are more than 10 times greater than those for January (Table 4). The contribution of wildfires causes July CO emissions to be higher than those in January by more than a factor of 100. Thus, natural VOC and CO emissions are nearly equal in January but CO emissions are about 3 times greater than VOCs in July.

4.6 Particulate matter

Natural particle emissions are divided into two categories: coarse and fine. Coarse mass (PM_c) is not speciated and is not chemically reactive in the CMAQ model (there is an exception to this – see below). However, fine mass (PM_{2.5}) is divided into chemical components and these are described separately.

**Fig. 6.** Total natural chlorine (HCl+ClNO₂+sea salt Cl) emissions, expressed as Mg of Cl, calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.**Table 3.** Monthly emissions totals for naturally occurring chlorine gases within the atmospheric modeling domain.

Species	Jan Mg	Jul Mg
HCl	1.7×10 ⁴	5.7×10 ³
ClNO ₂	8.6×10 ²	2.2×10 ¹
Total chlorine	1.7×10 ⁴	5.6×10 ³

Natural coarse mass originates primarily from windblown dust (anthropogenically-derived road dust is not considered “natural” in the current modeling) and wildfires. Thus, PM_c emissions are greater when and where winds are stronger and when/where fires occur. However, soil dryness and vegetation cover also play a role. Figure 8 illustrates the January and July spatial distribution of natural PM_c emissions. July PM_c emissions are roughly twice as high as those in January (Table 5). These PM_c values do not include coarse sea salt which is tracked separately in CMAQ. Coarse sea salt mass is confined to grid cells over the oceans and along the coasts. Total sea salt emissions (coarse and fine mass), listed separately in Table 5, are over four times greater in January than July because of stronger winter winds.

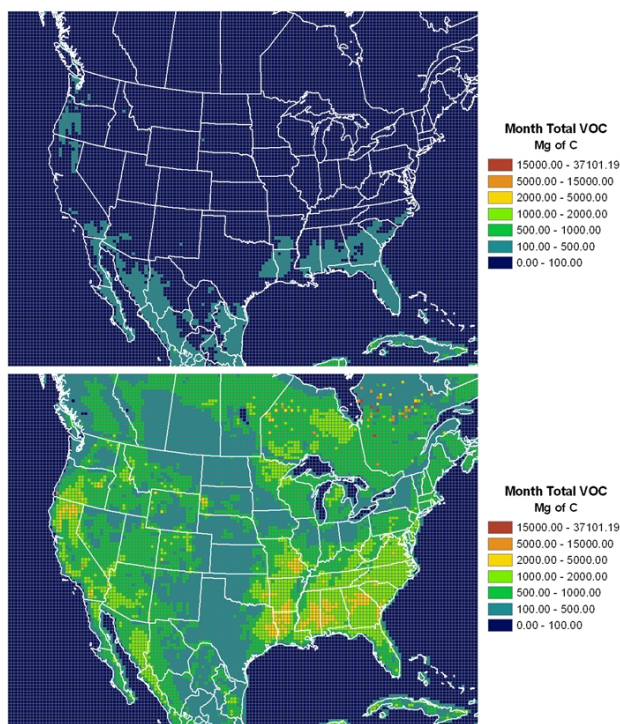


Fig. 7. Total natural VOC emissions (Mg of C) calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.

Table 4. Monthly emissions totals for naturally occurring non-methane volatile organic compounds (VOCs) and CO within the atmospheric modeling domain.

Species	Jan Mg	Jul Mg
VOCs ^a	5.3×10^5	7.7×10^6
CO	1.9×10^5	2.1×10^7
Total reactive carbon	6.1×10^5	1.7×10^7

^a Expressed as equivalent carbon.

Fine particle mass emissions from natural sources are divided into sulfate, nitrate, organic and elemental carbonaceous, nonreactive and sea salt aerosols. Sulfate, nitrate, OC, and EC are emitted by wildfires. The nonreactive component of PM_{2.5} is emitted by wildfires and windblown dust. Total PM_{2.5} mass emissions (which include the minor fine sea salt component) are plotted in Fig. 9 for January and July. In January PM_{2.5} is predominantly composed of nonreactive dust while OC is the major component in July (Table 5).

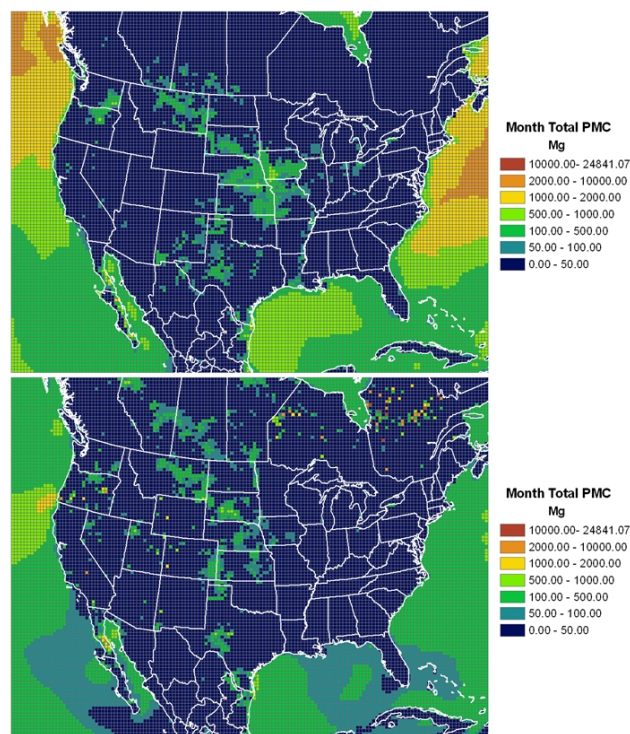


Fig. 8. Total natural PM_c emissions (Mg) calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.

Table 5. Monthly emissions totals for naturally occurring particulate matter within the atmospheric modeling domain.

Species	Jan Mg	Jul Mg
PM _c ^a	2.2×10^5	4.6×10^5
PM _{2.5} ^a	2.5×10^4	1.7×10^6
Sulfate ^b	2.7×10^1	3.3×10^4
Nitrate ^b	2.7×10^0	3.3×10^3
OC ^b	1.0×10^3	1.3×10^6
EC ^b	2.2×10^2	2.7×10^5
Nonreactive ^b	2.4×10^4	1.0×10^5
Sea salt ^c	4.5×10^6	1.1×10^6

^a Does not include sea salt component.

^b Fine mass only.

^c Includes both fine and coarse mass.

5 Comparisons of anthropogenic and natural emissions

Anthropogenic emissions that are compared here to natural emissions were compiled for the VISTAS regional haze modeling exercise (MACTEC, 2008). The VISTAS Base G2 emissions inventory is considered one of the best available for this period. Its spatial domain and 36-km grid structure is identical to that used in the current natural emissions

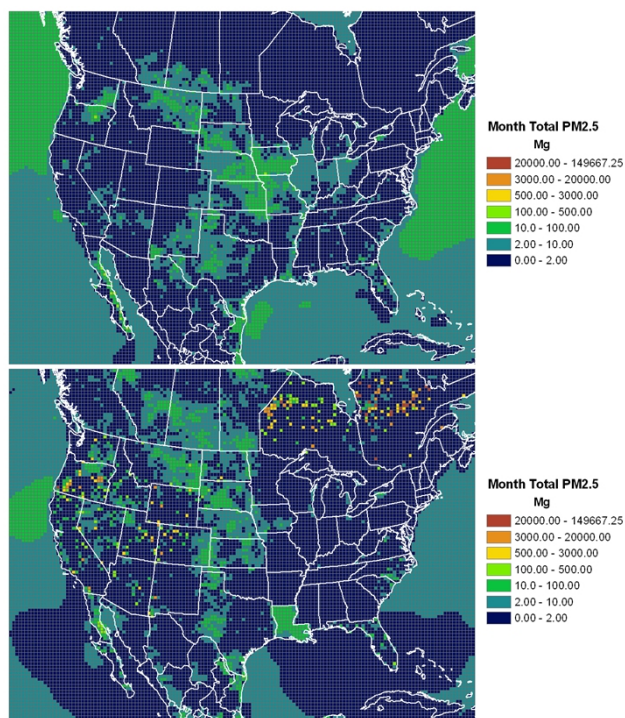


Fig. 9. Total natural $\text{PM}_{2.5}$ (sum of sulfate, nitrate, OC, EC, nonreactive and fine sea salt) emissions (Mg) calculated for January (top) and July (bottom) 2002 for the atmospheric modeling domain.

analysis. The Base G2 anthropogenic inventory represents “typical” 2000–2004 point source and US wildfire emissions in order to avoid the irregularities associated with both of these source classes. This assumption also affects the natural emissions data developed in this study because the wildfires emissions were based on the techniques and data supplied by VISTAS. However, wildfire emissions in Canada and Mexico have a more mixed pedigree and were not strictly representative of 2002 but the overall impact on results should be minor. Other than wildfires, the natural emissions described here are specific to the year 2002.

Figure 10 compares July VISTAS 2002 anthropogenic emissions with those in the enhanced July 2002 natural emissions data base developed for this effort. Total gaseous ($\text{SO}_2 + \text{DMS} + \text{H}_2\text{S}$) sulfur is dominated by anthropogenic emissions in 2002. The fraction attributable to man-made sources continues to decline with SO_2 emission controls on coal-fired electric power plants and low-sulfur diesel fuel for on-road use. However, anthropogenic sources will continue to dominate this emissions category over the next 10+ years.

Anthropogenic sources also make up the majority of NO_x emissions but only by about 12%. The natural emission fraction is reduced from 44% to about 21% if lightning NO_x is omitted from the data set. This omission is probably the most indefensible of the routine assumptions made when building emissions data sets for regulatory photochemical modeling.

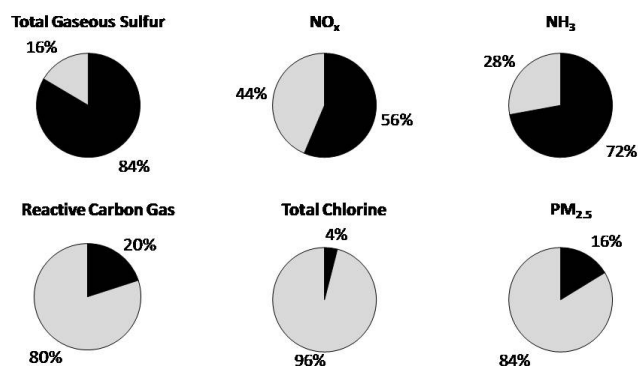


Fig. 10. Relative proportions of anthropogenic (solid black) and natural (gray) emissions within the modeling domain for the modified VISTAS July 2002 emissions inventory developed for this study. “Total Gaseous Sulfur” represents the sum of sulfur in $\text{SO}_2 + \text{DMS} + \text{H}_2\text{S}$. “Reactive Carbon Gas” represents the sum of carbon in non-methane VOCs+CO. “Total Chlorine” represents the sum of chlorine in $\text{HCl} + \text{ClNO}_2$ + sea salt. The $\text{PM}_{2.5}$ comparison includes sea salt in the fine particle fraction.

A future paper describes the impact of lightning NO_x on simulated air quality, including ground-level ozone.

Ammonia emissions are dominated by anthropogenic activities, especially livestock farming. The addition of ocean NH_3 emissions did little to alter the balance between natural and man-made NH_3 sources in this modeling domain.

No changes were made to emissions of VOCs or CO. The large dominance of natural sources emitting reactive carbonaceous gases is the same that has existed since photochemical modeling began over two decades ago. The dominance of natural emissions is not as large in winter due to the large decline in biogenic VOC emissions and wildfires.

Finally, about 99% of natural primary fine particle emissions are associated with windblown dust and wildfires. The recent addition of fine sea salt emissions did little to change this balance because sea salt is so geographically limited and most of its mass is in the coarse size fraction.

6 Conclusions

A wider range of natural pollutant emissions have been included in an expanded emissions data base for use in the US EPA CMAQ air regulatory model. Lightning NO_x , DMS and H_2S from oceans and continental sources, and chlorine (HCl and ClNO_2) from reactions involving sea salt are all new additions to the species already present in a data base developed for modeling regional levels of fine particles. Lightning NO_x in July was estimated to total 4.8×10^5 Mg equal to 52% of the total NO_x from all natural sources, 40% of all anthropogenic NO_x emissions, and 23% of the July sum of natural and anthropogenic NO_x emissions. This amounts to a large increase in total domain-wide NO_x emissions, consistent

with the findings of Kaynak et al. (2008). However, the characteristics of LNO_x are such that on average 90% of LNO_x is emitted above a height of 2500 m (i.e., well above the afternoon convective boundary layer). This means that LNO_x is unlikely to strongly influence ozone photochemistry in the vicinity of where it is generated and may be more important for its influence on background levels of NO_x and ozone in the free troposphere. Quantifying the implications for regional ozone and nitrogen deposition requires detailed model simulations.

Sources other than wildfires contribute significantly to the total of natural sulfur emissions. Even so, anthropogenic SO₂ emissions are still dominant but declining steadily. In winter, when fewer wildfires occur, geogenic and marine sources of sulfur account for nearly all natural sulfur emissions. In July, the sources added to the existing inventory account for 60% of all natural gaseous sulfur emissions. Taking advantage of the newly added emissions of DMS and H₂S require modifications to the gas and aqueous chemical modules in models like CMAQ. Impacts of these natural emissions on aerosol modeling are expected to be very small in polluted areas but should contribute to background particle levels in clean environments. Modeling the contributions of DMS and H₂S to air quality makes possible the use of models like CMAQ for characterizing natural aerosol background levels.

Recent experimental evidence and model results have demonstrated the importance of chlorine emissions on local photochemistry in polluted coastal environments (Knipping and Dabdub, 2003; Tanaka et al., 2003b; Chang and Allen, 2006). The addition of natural HCl and ClNO₂ emissions to an existing inventory increase total natural chlorine emissions by only a tiny amount (~1%). This is due to the predominance of sea salt chloride as the primary source of atmospheric chlorine. There is a need to update existing aerosol modules in models like CMAQ to more accurately simulate the dechlorination of sea salt particles and the subsequent release of reactive chlorine to the atmosphere. Changes are also required in CMAQ so that its chemistry modules can simulate oxidation reactions involving reactive chlorine. Many of the processes involved in emission production from natural sources, such as seasonal variance, still require further research. It is recommended that further investigation be performed on temporal allocation of the natural sources for which we applied default profiles. Furthermore, one caveat should be noted; there was a misinterpretation regarding diel release of oceanic ClNO₂ we present in Fig. A7 as referenced from Osthoff et al. (2008). Our interpreted temporal allocation fails to include nighttime levels provided in Osthoff et al. (2008) and most recently by Thornton et al. (2010). Fortunately our interpretation captures the early morning peak flux given by both Osthoff et al. (2008) and Thornton et al. (2010) and conserves diel totals. Only the diel variation is misconstrued; nonetheless, the emissions of ClNO₂ should receive further nighttime allocations in any future work as defined by studies such as Osthoff et al. (2008) and Thornton

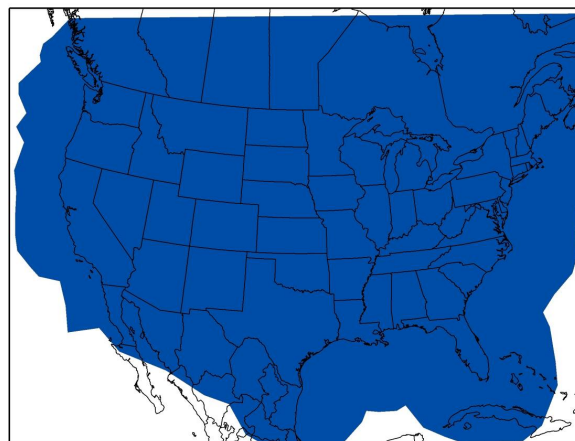


Fig. A1. The geographic extent of NLDN flash data included in this modeling is shown relative to the 36-km modeling domain.

et al. (2010). However, for our subsequent CMAQ modeling this oversight is very insignificant given that this GEIA inventory source of ClNO₂ represents less than 1% of the total chlorine source we simulated for our modeling inventory. Estimated uncertainties in emissions for natural sources range from about 50% to around a factor of 10 according to the sources we have referenced.

Appendix A

Supplemental information on the simulation of natural source emissions

A1 NO_x

Natural sources of NO_x simulated in this work include soils, wildfires, and lightning. Soil NO emissions were directly estimated using established methods within SMOKE via the BEIS (Biogenic Emissions Inventory System) model version 3.14. Emissions of NO_x from wildfires also used established methods by adopting the wildfire records in existing Regional Planning Organization (RPO) inventories for further processing within SMOKE. Lightning NO emissions from detected lightning flashes were based on lightning flash data as reported in 2002 by the National Lightning Detection Network (NLDN). The CMAQ modeling domain used in this work is the rectangular area covered by the map in Fig. A1. The spatial extent of the NLDN data set applied for this domain is shown in blue. Rather than applying a single NO production factor for flash emissions rates (as has been done elsewhere), the emissions rates for flashes in the 2002 simulation used an algorithm to estimate flash-specific emissions rates. This approach, referred to as the mean peak current method, is a variation on the methods described by Price et al. (1997), DeCaria et al. (2000), and Ott et al. (2010). The

basis of this approach is the cloud-to-ground (CG) flash data from the NLDN data set. We introduce separate treatment of positive flash return strokes and use all NLDN peak current values (rather than setting a minimum threshold) as recorded in the data set.

The average current (amperes) for both a negative (C_n) and a positive (C_p) CG flash was calculated as

$$C_n, C_p = F_t \cdot [S_1 + (F_s \cdot S_s)] \cdot I_{pc} \quad (\text{A1})$$

where F_t is the total number of flashes, S_1 and S_s are constants (3.12×10^{-4} and 1.39×10^{-4} , respectively, based on Price et al., 1997), F_s is the total number of return strokes of the flash minus the first return stroke, and I_{pc} is the peak current (A) for the flash.

The energy released in negative (E_n) and positive (E_p) CG flashes is calculated as

$$E_n = C_n \cdot V_n \quad (\text{A2})$$

and

$$E_p = C_p \cdot V_p \quad (\text{A3})$$

where V_n and V_p are the breakdown potentials in air for negative and positive flashes, respectively ($-3 \times 10^8 \text{V}$ and $5 \times 10^8 \text{V}$ based on Price et al., 1997). The total energy E_{cg} (in joules) per CG flash is then computed as

$$E_{cg} = E_n + E_p \quad (\text{A4})$$

Thus, the CG flash total NO was calculated as

$$T_{cg} = E_{cg} (P/N_a) \quad (\text{A5})$$

where P is the production rate of NO from a CG flash (a constant value of 1×10^{16} molecules per joule based on Price et al., 1997) and N_a is Avogadro's number needed to express T_{cg} in units of moles of NO.

Data on IC flashes are not included in NLDN data sets. Previous research on IC flashes suggested setting the IC/CG NO production ratio to 1/10. More recently, Ott et al. (2007, 2010) suggested a much stronger contribution from IC to NO production. After an extensive review of current literature and recent research we decided to set the IC NO production per flash equal to the CG NO production per flash primarily based on the research of Ott et al. (2007, 2010). An IC-to-CG flash ratio of 2.7 was assumed taken based on Pierce et al. (1999) which stems from the work of Price and Rind (1993). This falls in the range of climatological IC to CG ratios over the continental US (2.64–2.94) reported by Boccippio et al. (2001). It follows from this that IC NO production is computed as

$$T_{ic} = 2.7T_{cg} \quad (\text{A6})$$

The vertical atmospheric distribution of lightning NO_x has been studied by a number of researchers including Novak and Pierce (1993), Pickering et al. (1998), Wang et al. (1998),

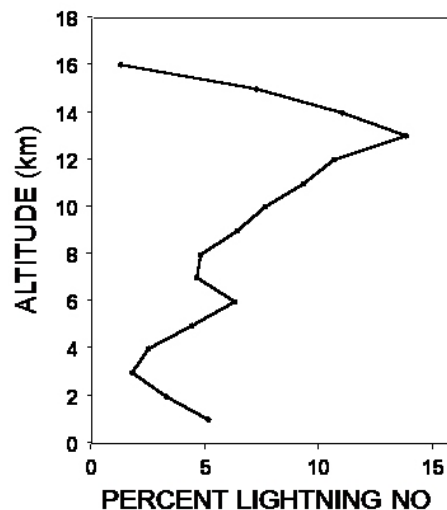


Fig. A2. Vertical allocation of lightning NO emissions based on modified Pickering et al. (1998) mid-latitude continental profile. Modifications were made as suggested by L. Ott (personal communication, 2007) to eliminate the large peak in the lowest kilometer in the Pickering et al. (1998) profile.

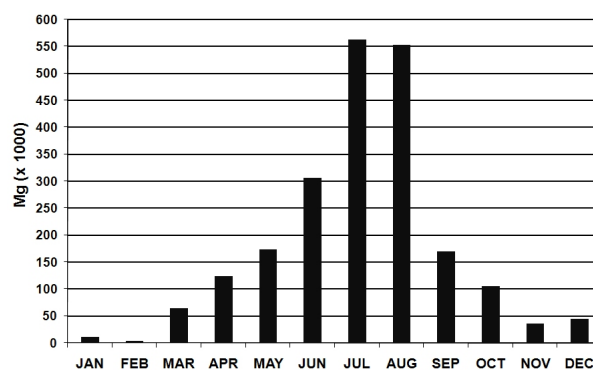


Fig. A3. Monthly domain-wide lightning NO total emissions for the 2002 natural emissions

DeCaria et al. (2000), Hudman et al. (2007), and Ott et al. (2007, 2010). Most of the published work uses some variation of the Pickering et al. (1998) vertical distribution. Thus, estimated lightning NO emissions in this analysis were distributed vertically through the troposphere using the mid-latitude continental profile from Pickering et al. (1998) modified as suggested by Lesley Ott (personal communication, 2007) to eliminate the large peak in the lowest kilometer. This profile, illustrated in Fig. A2, allocates about 5% of the NO to the lowest 1000 m of the atmosphere and redistributes the remaining mass to higher layers. Figure A3 illustrates the monthly distribution of estimated lightning NO emissions showing the expected strong summer peak.

Table A1. Natural source reduced sulfur emissions factors.

Source	H ₂ S μg m ⁻² day ⁻¹	DMS μg m ⁻² day ⁻¹
Ocean ^a	4	27–591 ^b
Freshwater Lakes ^c	64.8	38.4
Great Salt Lake ^d	64.8	43.4
Coastal Wetlands ^e	816	2664

^a H₂S factor is derived from Andreae and Jaeschke (1992).

^b Approximate factor range for ocean DMS based on reallocation of annual sulfur values from Kloster et al. (2006) to daily values.

^c Taken from values from the lower range of freshwater sulfur flux estimates derived from several studies including Richards et al. (1991), Bates et al. (1992), Giblin and Wieder (1992), Gröne (1997), Bodenbender et al. (1999), and Sharma et al. (1999).

^d Given the lack of available information on natural emissions from the Great Salt Lake the H₂S factor applied for freshwater lakes was conservatively applied. The DMS factor is derived from Richards et al. (1994).

^e Based on average of factors produced from a number of studies on sulfur flux from salt marsh and *Spartina Alterniflora* including Steudler et al. (1985), Giblin and Wieder (1992), and DeLaune et al. (2002).

Table A2. Sulfur emissions factors for geothermal sources.

Source (number of discrete emission points)	Sulfur Equivalent T/day
Mt. St. Helens (1) ^a	94
Mt. Hood (1) ^b	6
Mt. Baker (1) ^c	5.2
Yellowstone geothermal features (138) ^d	2.7 (for each)
Non-Yellowstone geothermal features (1703) ^e	0.1 (for each)

^a Approximated based on low-to-mid-range value extracted from SO₂ measurements from Chapter 26, Figure 16 of Sherrod et al. (2008).

^b From Chapter 26, Table 1 of Sherrod et al. (2008).

^c McGee et al. (2001).

^d William Evans, USGS, personal communication, 2005.

^e Set to 1% of the H₂S estimated for non-eruptive volcanoes by Berresheim and Jaeschke (1983).

A2 Sulfur

Natural sources of sulfur simulated in this work included SO₂ from wildfires, SO₂ and H₂S from geothermal sources, H₂S and DMS from inland lakes, oceans, coastal wetlands, and sea salt sulfate. Treatment of wildfire emissions in this work introduced no new techniques and simply used a top-down method of selecting wildfire records from the existing RPO inventories for further processing with the SMOKE system. Likewise, sulfates from sea salt required no unique treatment

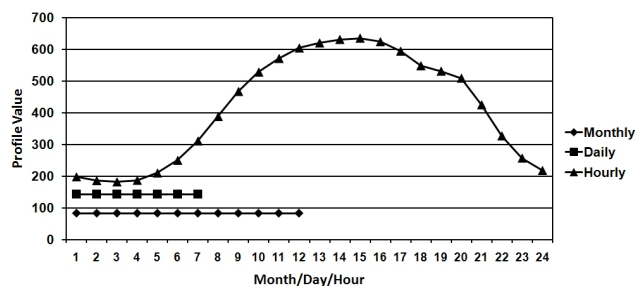


Fig. A4. SMOKE v2.5 default hourly temporal profile for natural sources applied to sulfur emissions from ocean, coastal wetland, freshwater, the Great Salt Lake, and geothermal sources. Default monthly and daily temporal allocation profiles are also shown.

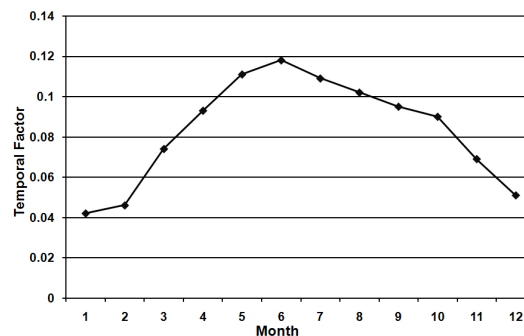


Fig. A5. Ocean DMS monthly temporal profile derived from the work of Kettle and Andreae (2000) applied to the annual mean flux of sulfur from the oceans as delineated by Kloster et al. (2006).

prior to processing in CMAQ because sea salt emissions are simulated from within CMAQ beginning with CMAQ version 4.5. The remaining natural sulfur sources required investigation of literature for acceptable emissions factors. Summaries of the emissions factors applied within this simulation are provided in Tables A1 and A2. Figure A4 illustrates the hourly temporal profile (the default within SMOKE v2.5) for natural sources that was applied in supplement to the emissions factors to produce hourly emissions for a representative day from oceans, coastal wetlands, freshwater lakes, the Great Salt Lake, and geothermal sources. The one instance in which direct allocation to a representative day was not employed involved DMS emissions from oceans. These emissions were first allocated into monthly values based on the work of Kettle and Andreae (2000) as shown in the monthly temporal profile provided in Fig. A5. Incorporating monthly/seasonal variability to the emissions from sulfur sources (other than ocean DMS) should be explored for future work.

Simulating reduced sulfur (H₂S and DMS) from soils was based on a simple proportionality factor. This factor was based on the ratio of the soil release of NO_x and reduced sulfur as estimated by NAPAP (1991). NAPAP (1991) provides

annual estimates of soil release of NO_x , H_2S , and DMS for the contiguous United States. A more sophisticated approach is not warranted at this time. Using NO_x as a spatial surrogate, the conversion factors derived from the NAPAP soil flux estimates were

$$\text{H}_2\text{S conversion factor} = 1,570 \text{ Mg} \quad (\text{A7})$$

of $\text{H}_2\text{S}/1,200,000 \text{ Mg}$ of NO_x

$$\text{DMS conversion factor} = 1,640 \text{ Mg} \quad (\text{A8})$$

of $\text{DMS}/1,200,000 \text{ Mg}$ of NO_x

The resulting conversion factors for H_2S (1.31×10^{-3}) and DMS (1.37×10^{-3}) were then directly applied to the SMOKE v2.5 Biogenic Emissions Inventory System (BEIS3.14) processor results for soil NO to estimate reduced sulfur emissions.

A3 Chlorine

Natural sources of chlorine simulated included sea salt aerosols of fine and coarse chloride (estimated from CMAQ as mentioned in the previous section on natural sulfur) and gaseous emissions of HCl and ClNO_2 from the oceans. The special treatment performed in this simulation to produce emissions of HCl and ClNO_2 is described further here. Annual totals of HCl and ClNO_2 from oceans were taken from the GEIA Reactive Chlorine Emissions Inventory (RCEI) (Erickson et al., 1999). This inventory is available online at <http://www.geiacenter.org/presentData/reactivechlorine.html>. The GEIA RCEI inventory was then spatially allocated to our modeling domain using GIS (geographic information system) software. The GEIA RCEI provided annual totals for the GEIA $1^\circ \times 1^\circ$ global domain that were then assigned to the $36 \times 36 \text{ km}$ grid cells in the modeling domain. Once the annual totals per grid cell were calculated they were then allocated into natural and anthropogenic components. The GEIA data set did not provide separate anthropogenic and natural emission estimates. Therefore, factors had to be produced to create these two separate components. This was done by determining the ratio of natural to anthropogenic NO_x over land near the oceans and then partitioning the GEIA fluxes proportionately into natural and anthropogenic components. NO_x ratios for natural and anthropogenic component differentiation were created for each month from the emission data base. Essentially, a natural weighting or percentage was created with the remainder assumed to be anthropogenic. The natural and anthropogenic component ratios calculated are provided in Table A3. Before these monthly natural and anthropogenic component split factors could be applied, the annual GEIA RCEI emissions totals per modeling grid cell had to first be allocated to each month of the simulation. Table A4 provides the allocations factors for distributing annual totals to

Table A3. Monthly split-factors (component percentages) derived from ocean/coastline NO_x ratios that were applied to the GEIA RCEI (Reactive Chlorine Emissions Inventory) to produce natural and anthropogenic components from annual total chlorine (i.e. HCl and ClNO_2 referenced in Table A4).

Month	NATURAL % of Total Chlorine	ANTHROPOGENIC % of Total Chlorine
Jan	13	87
Feb	12	88
Mar	14	86
Apr	16	84
May	14	86
Jun	16	84
Jul	17	83
Aug	17	83
Sep	16	84
Oct	16	84
Nov	16	84
Dec	14	86

Table A4. Monthly allocation factors applied to the GEIA RCEI (Reactive Chlorine Emissions Inventory) to produce monthly HCl and ClNO_2 emissions from annual totals. Factors in this table may not sum to one due to rounding.

Month	HCl	ClNO_2
Jan	.153198	.26936
Feb	.157936	.16835
Mar	.125822	.06734
Apr	.089628	.03367
May	.049091	.010101
Jun	.041326	.006734
Jul	.038562	.005051
Aug	.039747	.005051
Sep	.047380	.013468
Oct	.055146	.050505
Nov	.088575	.13468
Dec	.113582	.23569

monthly values. The input used to calculate these annual-to-monthly allocation factors were taken from Erickson et al. (1999).

Finally, monthly total emissions per 36-km grid cell were allocated into a single representative day of emissions for each month with hourly values temporally allocated based on Keene et al. (2007) and Osthoff et al. (2008). The temporal allocation profile applied for HCl is shown in Fig. A6 while the profile applied for ClNO_2 is shown in Fig. A7 (these profiles were created by visual approximation from figures provided in the sources noted). The representative day of emissions created for each month was then used to represent each remaining simulation.

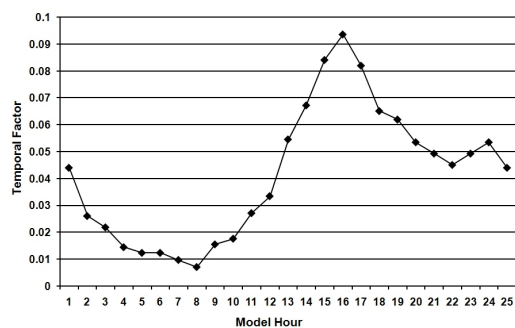


Fig. A6. Hourly temporal profile derived from the diel variability of HCl provided in Keene et al. (2007). This was applied to perform hourly allocations for the representative day of emissions created for oceanic HCl.

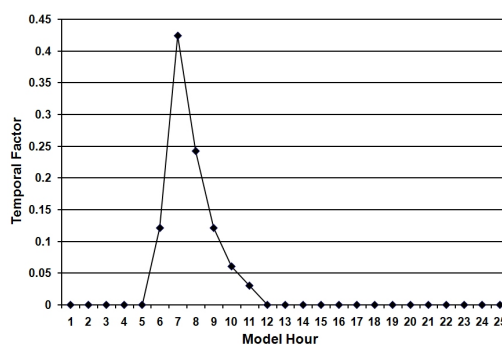


Fig. A7. Hourly temporal profile derived from the diel variability of ClNO₂ provided in Osthoff et al. (2008). This was applied to perform hourly allocations for the representative day of emissions created for oceanic ClNO₂. NOTE: See caveat mentioned in our conclusions.

A4 NH₃

Natural sources of ammonia included in this simulation were oceans, wild animals, and wildfires. The role of soils as a source of NH₃ is described in the main paper along with the methodology for treating wildfires. The method of estimating natural emissions of NH₃ from wild animals was based on the work described by Mansell (2005). The CMU Ammonia Model version 3.6 (Davidson et al., 2004) was used to estimate the total wild animal NH₃ emissions from the contiguous US. The CMU model was ran with the inherent activity data (used “as-is” with no modifications or additional quality assurance) and emission factors for bears, deer, and elk. Emissions factors from the CMU model are provided in Table A5. The resulting annual total NH₃ (kg) emissions by county from the CMU Model run were then spatially allocated to counties using a GIS so that subsequent allocations to model grid cells could be performed. The wild animal emissions were considered temporally invariant in our work in agreement with work by both Davidson et al. (2004) and Mansell (2005). The guidance taken from Mansell (2005) used forest, grass, and shrubland to spatially allocate the wild animal NH₃. However, given that our modeling assumes only natural land cover exists within the modeling domain, spatial allocation to these three types was not performed. Instead, county level resolution emissions were deemed sufficient and the annual total emissions were allocated based on the fraction of county in each modeling grid cell. These gridded annual emissions were subsequently converted to an hourly emissions file for one representative day with no hourly variation, consistent with the temporal invariance assumption. Subsequently, a prototypical day was replicated to represent all days of the year.

Canadian wild animal emissions were estimated slightly differently. We determined animal populations (activity) and their native ranges, known habitat boundaries, or areas considered as wild land areas suitable for habitation and then allocated these populations to the habitats. These values

Table A5. CMU Ammonia Model v3.6 wild animal emissions factors.

Source	Emission Factor (kg NH ₃ per animal yr ⁻¹)
Bear (Black & Grizzly)	0.378
Deer	0.378
Elk	2.04

were allocated to the modeling grid cells using GIS software. Animal population emissions factors were temporally allocated in the same invariant manner as in the US. Canadian activity data for whitetail and mule deer, black and grizzly bear, elk, and moose were acquired for British Columbia from the British Columbia Ministry of Environment, Lands and Parks 2000–2001 publications. The CMU Model emission factor for elk (Table A5) was used as the surrogate emission factor for moose. Activity data for the remaining Canadian provinces within our modeling domain (Alberta, Saskatchewan, Manitoba, Ontario, Quebec, New Brunswick, and Nova Scotia) were not as well documented. However, estimated populations were obtained as possible and are shown by Canadian province in Table A6. A habitat surrogate factor was created for these remaining provinces based on the BELD3 aggregated allocations of forest, grassland, and shrubland for our modeling domain grid cells. This habitat (i.e. wild land) surrogate factor was applied to the species-specific native range to allocate animal populations within this native range to each modeling grid cell. Native ranges for the wild animals modeled came from Patterson et al. (2007) except for the elk range in Manitoba which was available from the Manitoba Wildlife and Ecosystem Protection Branch over the Internet (<http://www.gov.mb.ca/conservation/wildlife/mbsp/fs/elk.html>).

There is limited information on Mexican animal populations. Thus, we used the GEIA wild animal NH₃ emissions

Table A6. Populations of Canadian wild animals^a.

Province	White-tail Deer	Mule Deer	Moose	Elk	Black Bear	Grizzly Bear
Alberta ^b	232,000	133,000	118,000	26,000	40,000	500
Saskatchewan ^c	369,263	43,028	43,196	14,782	35,000	N/A
Manitoba ^d	250,000 ^e	N/A	32,000	7,000	27,500	N/A
Ontario ^{f,g}	400,000	N/A	200,000	N/A	70,000	N/A
Quebec ^{g,h}	285,000 ⁱ	N/A	125,000	N/A	60,000	N/A
New Brunswick ^{j,k}	85,000	N/A	25,000	N/A	12,500	N/A
Nova Scotia ^{g,l}	60,000	N/A	9,000	N/A	3,000	N/A

^a Caribou, antelope, and other ungulate populations are not included in the population estimates. N/A=population estimates not acquired/available at the time of this modeling or data not found for native ranges.

^b Government of Alberta Sustainable Resource Development website has since been modified; however, original source is archived and viewable at: <http://web.archive.org/web/20080603222931/www.srd.gov.ab.ca/fishwildlife/livingwith/huntingalberta/gamespecies.aspx>, <http://web.archive.org/web/20071103182904/www.srd.gov.ab.ca/fishwildlife/wildlifeinalberta/bearsalberta/presentstatus.aspx>, accessed: 12 January 2010.

^c Status and management of Wildlife in Saskatchewan: 2002 and 2003 Resource Technical Report: 2005-2. Saskatchewan Environment Resource Stewardship Branch website has since been modified; however, original source is archived and viewable at: <http://web.archive.org/web/20060212185403/www.se.gov.sk.ca/fishwild/Status+of+Wildlife+2002-03.pdf>, accessed: 12 January 2010.

^d Manitoba Conservation Wildlife and Ecosystem Protection Branch website has since been modified; however, original source is archived and viewable at: <http://web.archive.org/web/20050321140253/www.gov.mb.ca/conservation/wildlife/managing/>, accessed: 12 January 2010.

^e 1995 historic peak.

^f Ontario Natural Resources Information Center (2006).

^g US Geological Survey Biology website has since been modified; however, original source is archived and viewable at: <http://web.archive.org/web/20070715151620/http://biology.usgs.gov/s+t/noframe/c286.htm>, accessed: 12 January 2010.

^h Quebec Ministry for the Natural Resources and Fauna (2006).

ⁱ Mainland population.

^j New Brunswick Department of Natural Resources (2006).

^k North American Bear Center website has since been modified; however, original source is archived and viewable at: <http://web.archive.org/web/20071121142706/http://www.bear.org/Black/BlackBearPopulations.html>, accessed: 12 January 2010.

^l Nova Scotia Wildlife Division (2006).

data set spatially allocating the GEIA $1^\circ \times 1^\circ$ global dataset to the 36×36 km resolution grid cells. Wild animal NH_3 emissions provided within the GEIA database are actually based on the 1990 estimates from the EDGAR 2.0 database. Annual totals from the GEIA grid were allocated to the modeling domain and handled in a similar fashion to that done for US wild animal emissions.

Natural emissions of NH_3 from ocean surfaces within the modeling domain were included based on annual ocean NH_3 emissions estimates from the GEIA database. These were allocated to grid cells based on percent area of ocean per grid cell using a GIS. Emissions were directly allocated temporally from annual values to a daily value assuming no seasonal differences. This seasonal invariance is based on the SMOKE v2.5 monthly temporal profile for salt water allocation. Hourly temporal variation was derived by applying the SMOKE v2.5 temporal profile for salt water to the daily total emissions. Temporal profiles are shown in Fig. A4. Ocean NH_3 emissions were created for a single representative day and this representative day was used for the remaining days of the year.

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