Home

Search Collections Journals About Contact us My IOPscience

Source attribution of particulate matter pollution over North China with the adjoint method

This content has been downloaded from IOPscience. Please scroll down to see the full text.

2015 Environ. Res. Lett. 10 084011

(http://iopscience.iop.org/1748-9326/10/8/084011)

View the table of contents for this issue, or go to the journal homepage for more

Download details:

IP Address: 210.77.64.109 This content was downloaded on 13/04/2017 at 06:23

Please note that terms and conditions apply.

You may also be interested in:

Anthropogenic fugitive, combustion and industrial dust is a significant, underrepresented fine particulate matter source in global atmospheric models Sajeev Philip, Randall V Martin, Graydon Snider et al.

Impact of aerosol–meteorology interactions on fine particle pollution during China's severe haze episode in January 2013 Jiandong Wang, Shuxiao Wang, Jingkun Jiang et al.

Exceedances of air quality standard level of PM2.5 in Japan caused by Siberian wildfires Kohei Ikeda and Hiroshi Tanimoto

Environmental effects of the recent emission changes in China: implications for particulate matter pollution and soil acidification Bin Zhao, Shuxiao Wang, Xinyi Dong et al.

Satellite detection and model verification of NOx emissions from power plants in NorthernChina Siwen Wang, David G Streets, Qiang Zhang et al.

Recent reduction in NOx emissions over China: synthesis of satellite observations and emission inventories

Fei Liu, Qiang Zhang, Ronald J van der A et al.

Public health impacts of the severe haze in Equatorial Asia in September–October 2015:

demonstration of a new framework for informing fire management strategies to reduce downwind smoke exposure

Shannon N Koplitz, Loretta J Mickley, Miriam E Marlier et al.

# **Environmental Research Letters**

# LETTER

CrossMark

OPEN ACCESS

RECEIVED 14 May 2015

REVISED

1 July 2015

10 July 2015

PUBLISHED 12 August 2015

Content from this work may be used under the terms of the Creative

Commons Attribution 3.0 licence.

Any further distribution of this work must maintain attribution to the author(s) and the title of the work, journal citation and DOI.



Source attribution of particulate matter pollution over North China with the adjoint method

# Lin Zhang<sup>1</sup>, Licheng Liu<sup>1</sup>, Yuanhong Zhao<sup>1</sup>, Sunling Gong<sup>2</sup>, Xiaoye Zhang<sup>2</sup>, Daven K Henze<sup>3</sup>, Shannon L Capps<sup>3</sup>, Tzung-May Fu<sup>1</sup>, Qiang Zhang<sup>4</sup> and Yuxuan Wang<sup>4,5,6</sup>

- Laboratory for Climate and Ocean-Atmosphere Sciences, Department of Atmospheric and Oceanic Sciences, School of Physics, Peking University, Beijing 100871, People's Republic of China
- <sup>2</sup> Key Laboratory for Atmospheric Chemistry, Chinese Academy of Meteorological Sciences, CMA, Beijing, People's Republic of China
- <sup>3</sup> Department of Mechanical Engineering, University of Colorado, Boulder, CO 80309, USA
- <sup>4</sup> Ministry of Education Key Laboratory for Earth System Modeling, Center for Earth System Science, Tsinghua University, Beijing 100084, People's Republic of China
- Department of Marine Sciences, Texas A&M University at Galveston, Galveston, TX 77553, USA
- <sup>6</sup> Department of Atmospheric Sciences, Texas A&M University, College Station, TX 77843, USA

<sup>0</sup> E-mail: zhanglg@pku.edu.cn

Keywords: source attribution, particulate matter, PM2.5, adjoint, pollution in China, regional transport

#### Abstract

We quantify the source contributions to surface PM2.5 (fine particulate matter) pollution over North China from January 2013 to 2015 using the GEOS-Chem chemical transport model and its adjoint with improved model horizontal resolution  $(1/4^{\circ} \times 5/16^{\circ})$  and aqueous-phase chemistry for sulfate production. The adjoint method attributes the PM2.5 pollution to emissions from different source sectors and chemical species at the model resolution. Wintertime surface PM2.5 over Beijing is contributed by emissions of organic carbon (27% of the total source contribution), anthropogenic fine dust (27%), and SO<sub>2</sub> (14%), which are mainly from residential and industrial sources, followed by NH<sub>3</sub> (13%) primarily from agricultural activities. About half of the Beijing pollution originates from sources outside of the city municipality. Adjoint analyses for other cities in North China all show significant regional pollution transport, supporting a joint regional control policy for effectively mitigating the PM2.5 air pollution.

# 1. Introduction

Severe particulate matter pollution in recent years has become a major environmental problem in China and has received broad interest worldwide (van Donkelaar *et al* 2010, Zhang *et al* 2013, He *et al* 2014, Guo *et al* 2014, Huang *et al* 2014). In January 2013, northern and eastern China experienced record-high polluted haze events with daily mean PM2.5 concentrations exceeding 500  $\mu$ g m<sup>-3</sup> at some cities in North China, raising deep public concerns on the health effects, particularly in the populated Beijing–Tianjin– Hebei region (Wang *et al* 2014a, 2014b). Better understanding of the sources and processes controlling the PM2.5 pollution over North China has become an important and urgent issue with implications for air quality policy.

PM2.5 includes various directly emitted primary aerosols and secondary aerosols that are formed through chemistry of gaseous precursors in the atmosphere (Seinfeld and Pandis 2006). The complexity in composition and nonlinearity in chemistry make it a challenge to quantify the contributing sources for PM2.5 pollution. Previous studies on source attribution of Chinese air pollution are based on measurement statistics (such as positive matrix factorization (PMF)) (Zhang et al 2013, Tao et al 2014, Zhang et al 2014a), back-trajectory receptor models (Wang et al 2004, Zhang et al 2013), the response surface modeling technique (Wang et al 2011) or sensitivity simulations with a chemical transport model (CTM) (Wang et al 2014a). However, these methods generally ignore the nonlinear chemistry of aerosol formation, or require unaffordable computational time. Here we **IOP** Publishing

use the nested-grid GEOS-Chem CTM and its adjoint model with a horizontal resolution of  $1/4^{\circ} \times 5/16^{\circ}$  to examine the sources of PM2.5 pollution over North China, and we interpret the source attribution results based on emission sectors, chemical species, and local versus regional transport influences.

### 2. Model description and method

The GEOS-Chem CTM (http://geos-chem.org, Bey *et al* 2001) is driven by assimilated meteorological data from the Goddard Earth Observing System (GEOS) of the NASA Global Modeling and Assimilation Office (GMAO). The latest version of GEOS-FP dataset has a temporal resolution of 3 h (1 h for surface variables and mixing depths) and a horizontal resolution of 1/ $4^{\circ} \times 5/16^{\circ}$ . We use a nested-grid version of GEOS-Chem (Chen *et al* 2009, Zhang *et al* 2014b) with the native 1/ $4^{\circ} \times 5/16^{\circ}$  horizontal resolution over East Asia (70°E–140°E, 15°N–55°N) and 2° × 2.5° over the rest of the world.

The model simulates a fully coupled tropospheric ozone- $NO_x$ -hydrocarbon-aerosol chemistry (Park *et al* 2004), with gas-aerosol phase partitioning of the ammonium- sulfate–nitrate–water system calculated with the ISORROPIA II thermodynamic equilibrium model (Fountoukis and Nenes 2007). Details of the model chemical scheme and global emissions have been described by Zhang *et al* (2012, 2014b) in the context of nitrogen deposition and surface ozone pollution over the United States. Model simulated PM2.5 concentrations are sums of black carbon (BC), organic carbon (OC), sulfate ( $SO_4^{2-}$ ), nitrate ( $NO_3^{-}$ ), ammonium ( $NH_4^{+}$ ), and fine dust concentrations.

For anthropogenic emissions over China we use the Multi-Resolution Emission Inventory of China for the year 2010 (MEIC, http://meicmodel.org), which is an updated version of the bottom-up emission inventory developed by Zhang et al (2009) and Lei et al (2011). Emissions are separated to five major sectors including industry, residential, power plants, transportation, and agriculture (Zhang et al 2009). Primary aerosols and gas precursors emissions over China for January are estimated to be:  $SO_2$  (2.6 Tg),  $NO_x$  (2.4 Tg as NO<sub>2</sub>), NH<sub>3</sub> (0.55 Tg), BC (0.23 Tg), OC (0.62 Tg), and anthropogenic fine dust (0.62 Tg). The standard GEOS-Chem model only includes emissions of natural mineral dust in four size bins with radii 0.1-1.0, 1.0-1.8, 1.8-3.0, and 3.0-6.0 µm as described by Fairlie et al (2007). They are involved in tropospheric chemistry through aerosol effects on photolysis rates (Martin et al 2003) and heterogeneous aerosol chemistry parameterized as uptake coefficients (Jacob 2000, Fairlie et al 2010). Bottom-up emission estimates suggest that other anthropogenic primary PM2.5 emissions besides BC and OC (mainly Ca and Mg) over China can be significant (Zhang et al 2007, Lei et al 2011). Here we have implemented in GEOS-

Chem the anthropogenic PM2.5 emissions (Lei *et al* 2011) and treated them as the finest mineral dust.

The adjoint model offers a computationally efficient approach to calculate the sensitivity of a model scalar (e.g. PM2.5 concentration at a given receptor) to all model parameters (e.g. emissions in all model grid cells) in a single model run. As described in Henze et al (2007, 2009) and Zhao et al (2015), the GEOS-Chem model can be viewed as a numerical operator F:  $\mathbf{c}_{n+1} = \mathbf{F}(\mathbf{c}_n, \mathbf{p})$ , where  $\mathbf{c}_n$  denotes all tracer concentrations at time step n and p is a vector of model parameters such as emissions. Considering the PM2.5 concentration at a given receptor and time, J, we define the sensitivity  $\lambda_p^n = \left(\frac{\partial J}{\partial p_n}\right)^{\mathrm{T}}$  and  $\lambda_c^n = \left(\frac{\partial J}{\partial c_n}\right)^{\mathrm{T}}$ , then  $\lambda_p^0 = \nabla_p J$  represents the sensitivity of J to model parameters (meaning how much J would change due to small perturbations in **p**), and  $\lambda_c^0 = \nabla_{c0} J$  represents its sensitivity to the initial conditions. The adjoint model iteratively computes those sensitivities simultaneously backwards in time:  $\lambda_c^{n-1} =$  $\left(\frac{\partial F}{\partial c}(\mathbf{c}_{n-1}, \mathbf{p})\right)^{\mathrm{T}} \lambda_{c}^{n}$  and  $\lambda_{p}^{n-1} = \left(\frac{\partial F}{\partial p}(\mathbf{c}_{n-1}, \mathbf{p})\right)^{\mathrm{T}}$  $\lambda_{c}^{n} + \lambda_{p}^{n}$ , where  $\left(\frac{\partial F}{\partial c_{n}}\right)^{T}$  and  $\left(\frac{\partial F}{\partial p}\right)^{T}$  are the transpose of the model Jacobian matrices with respect to  $c_n$  and p.

The adjoint of GEOS-Chem was developed by Henze et al (2007) and Kopacz et al (2009) for optimizing aerosol and CO sources, respectively. The adjoint includes transport, gas-phase chemistry, and heterogeneous chemistry to fully represent the inorganic PM2.5 simulation (Henze et al 2007, 2009). Capps et al (2012) further implemented the adjoint of the ISORROPIA II model. The GEOS-Chem adjoint model has been validated and applied in a number of studies to analyze aerosol sensitivities and optimize aerosol sources (Henze et al 2007, 2009, Wang et al 2012, Kharol et al 2013, Zhu et al 2013). We extend here the capability of the GEOS-Chem adjoint to a nested-grid version driven by the GEOS-FP data with the fine  $1/4^{\circ} \times 5/16^{\circ}$  horizontal resolution. This fine resolution allows GEOS-Chem to better represent the heterogeneity in topography and emissions over North China (Chen et al 2009), which is particularly important for resolving urban air pollution discussed in this study.

#### 3. Results and discussion

#### 3.1. Model simulation of PM2.5 over China

We compare the model simulation of PM2.5 concentration and chemical composition to surface measurements by the China Meteorological Administration (CMA) for the winter month of January. A number of studies have analyzed the episodic haze events during January 2013 (He *et al* 2014, Wang *et al* 2014b, Zhang *et al* 2014a). We focus on the monthly averages representing the mean atmospheric condition to target



emission control strategies for a long-term air quality improvement. Previous studies have shown that the haze pollution events in January 2013 over North China were characterized by large amounts of sulfate aerosol and models underestimated the observed sulfate enhancements likely due to missing mechanisms of heterogeneous chemistry (He et al 2014, Wang et al 2014b, Zheng et al 2015). The standard GEOS-Chem model simulates sulfate production from aqueous-phase oxidation of SO<sub>2</sub> by hydrogen peroxide (H<sub>2</sub>O<sub>2</sub>) and ozone (O<sub>3</sub>) in clouds, and gas-phase oxidization of SO<sub>2</sub> by OH (Park et al 2004). By adding heterogeneous uptake of SO2 on deliquesced aerosols under high-relative humidity conditions, Wang et al (2014b) partly improved the model sulfate simulation. In this study we examine the mechanism of aqueousphase oxidation of S(IV) (the sum of dissolved SO<sub>2</sub>,  $HSO_3^-$ , and  $SO_3^{2-}$ ) by dissolved nitrogen dioxide (NO<sub>2</sub>) that has been documented in the literature (Lee and Schwartz 1983, Clifton et al 1988, Sarwar et al 2013) but not yet included in GEOS-Chem.

The mechanism proposed by Lee and Schwartz (1983)gives the aqueous-phase reaction:  $\mathrm{HSO}_3^- + 2\mathrm{NO}_2 + \mathrm{H}_2\mathrm{O} \rightarrow \mathrm{SO}_4^{2-} + 3\mathrm{H}^+ + 2\mathrm{NO}_2^-.$ Clifton et al (1988) further studied the reaction and found rate constants 1-2 orders of magnitude higher than Lee and Schwartz (1983). The rate constants were summarized by Sarwar et al (2013) as a function of  $k = -7.1557 \times 10^4 \text{pH}^2 + 5.0 \times 10^6 \text{pH}.$ pH: Although NO<sub>2</sub> is only weakly soluble, the reaction can be important under conditions with high NO<sub>2</sub> concentrations and pH values (Littlejohn et al 1993). We implement this aqueous-phase reaction on both cloud droplets and deliquescent aerosols using the rate constant function as described above. The pH values are assumed to be 4.5 in clouds (Alexander et al 2009) and

5.6 for deliquescent aerosols based on aerosol measurements taken at a suburban site near Beijing in winter 2012 (T-M Fu, unpublished data). Figure 1 shows the simulated mean sulfate aerosol concentrations in surface air over North China for January 2013 from the standard and improved simulations. With this reaction, the simulation shows on average 68% increases in January mean sulfate concentrations at the surface of North China. Using a pH value of 4.5 for deliquescent aerosols would decrease the sulfate concentrations by 14% relative to the results with pH value of 5.6. In addition, we apply the aqueousphase oxidation of SO<sub>2</sub> by H<sub>2</sub>O<sub>2</sub> and O<sub>3</sub> on deliquescent aerosols, further increasing the surface sulfate North concentrations over China by 22% (figure 1(c)).

Figure 2 compares monthly mean observed versus simulated total PM2.5 concentrations at 34 CMA monitoring sites for January 2013. Aerosol chemical composition measurements including sulfate, nitrate, ammonium, black carbon (BC), and organic carbon (OC) at 13 CMA monitoring sites for January 2006–2007 (Zhang *et al* 2012) are also shown in figure 2. The model generally captures the spatial distributions of measured PM2.5 and aerosol composition concentrations with correlation coefficients of 0.65–0.84 and small biases for sulfate, nitrate and ammonium (within 10%; up to -23% for BC).

The model underestimates OC concentrations by 43% on average, particularly over the sites in the western China and southern China. This large OC underestimate has also been discussed by Fu *et al* (2012). Using surface measurements including those presented here, they pointed out a national-scale underestimate of the Chinese OC anthropogenic emissions



Figure 2. Model simulation of surface concentrations of PM2.5 and aerosol composition including sulfate ( $SO_4^-$ ), intrate ( $NO_3$ ), ammonium ( $NH_4^+$ ), black carbon (BC) and organic carbon (OC) in January over China. Measurements of PM2.5 (January 2013) and aerosol composition (January 2006–2007) at the China Meteorological Administration (CMA) surface sites (circles) are compared with model results (January 2013). Mean observed and simulated concentrations, and correlation coefficients are shown inset.

and in western China a missing OC source associated with biofuels for heating during the cold months. In addition, our study does not include secondary organic carbon, which accounts for 21% of the annual mean surface OC in China as simulated by GEOS-Chem (Fu *et al* 2012).

The model successfully simulates the high sulfate concentrations over North China in winter owing to the improved aqueous-phase chemistry for sulfate production as described above. The finer model horizontal resolution also improves the comparison by better capturing the spatial variability in PM distributions. We find BC and OC concentrations in a simulation at the coarser  $2^{\circ} \times 2.5^{\circ}$  resolution are respectively 48% and 65% lower than the measurements.

#### 3.2. Adjoint source attribution for PM2.5 pollution

We use the adjoint model to calculate the sensitivity of surface PM2.5 concentration over Beijing (the 1/ $4^{\circ} \times 5/16^{\circ}$  grid cell covering the center of Beijing: 39.9° N, 116.3°E) to all aerosol sources over North China. Figure 3 shows the adjoint sensitivities for the mean surface PM2.5 concentration at Beijing averaged from January 2013 to 2015 separated into different source sectors and chemical species at the model underlying resolution. The magnitudes represent approximately how much the January mean PM2.5 at Beijing would reduce in the absence of the specific emissions. Summing all values over the model domain gives  $122 \,\mu \text{g m}^{-3}$ , about 86% of the simulated mean PM2.5 concentration ( $142 \,\mu \text{g m}^{-3}$ ). The discrepancy is



mainly attributed to the nonlinear response of PM2.5 to emissions (Henze *et al* 2007, 2009). We find contributions from natural emissions (e.g. NH<sub>3</sub> from oceans and NO<sub>x</sub> from soils) and sources outside the model domain are small (less than 2%).

Figure 3 shows large source contributions from residential and industrial emissions of SO<sub>2</sub>, OC, and anthropogenic fine dust, consistent with the intensified use of coal over North China during the heating season (Zhang *et al* 2014a, Zhang *et al* 2013). Large source contributions are also seen from agricultural emissions of NH<sub>3</sub> reflecting formation of ammonium sulfate and ammonium nitrate aerosols (Wang *et al* 2011, Kharol *et al* 2013). We find in the model increasing NO<sub>x</sub> emissions would enhance the wintertime ozone titration, and thus suppress sulfate production from SO<sub>2</sub> oxidation by ozone, leading to some negative sensitivity with respect to NO<sub>x</sub> emissions.

We sum the adjoint sensitivities across different dimensions and interpret them as percentage contributions to the total sensitivity. By chemical species, the January mean PM2.5 over Beijing is generated by emissions of OC (27.3%), anthropogenic fine dust (27.2%), SO<sub>2</sub> (14.2%), NH<sub>3</sub> (13.2%), BC (9.6%), and NO<sub>x</sub> (8.5%). Zhang *et al* (2013) also showed largest fractions of OC (28.6%) and dust (28.9%) in

wintertime PM2.5 measurements in Beijing. Previous studies suggested that secondary inorganic aerosols (sulfate, ammonium, and nitrate) over North China are most sensitive to NH<sub>3</sub> emissions and least to NO<sub>x</sub> emissions in winter (Kharol *et al* 2013, Wang *et al* 2013). Our results also show weakest sensitivity for NO<sub>x</sub> emissions, but higher sensitivity for SO<sub>2</sub> than NH<sub>3</sub> due to the enhanced aqueous-phase sulfate production.

By emission sector, residential (49.8%) and industrial sources (26.5%) show the largest sensitivities, followed by agriculture (11.8%), transportation (5.9%), and power plants (6%). The estimated contributions heavily rely on the bottom-up sectorial emissions with uncertainties in both the emission magnitude and sectorial contribution. The contribution from transportation can be largely underestimated because secondary organic aerosols (SOA) are not included. Recent laboratory studies show emissions from gasoline vehicles cause significant formation of SOA that can be 6-10 times higher than their primary OC emissions (Nordin et al 2013, Gordon et al 2014). In a sensitivity simulation with OC emissions from transportation increased by a factor of 10, we find that the resulting contribution of transportation can reach 12%.



as labeled in the top-left panel. The open circles denote locations of the three cities.

Comparison with previous studies using the PMF method (Zhang et al 2013, Huang et al 2014) is complicated by the different source categories identified. Huang et al (2014) found dominant contributions from coal burning and secondary aerosols for 5-25 January 2013. Zhang et al (2013) estimated major source contributions from coal burning (57%), soil dust (16%), industrial pollution (12%), and secondary inorganic aerosol (6%) for PM2.5 over Beijing in winter 2010. Our adjoint results are qualitatively consistent with the PMF results; both show the largest contribution from coal burning, which dominates the residential and industrial emissions. The PMF results also indicated the importance of secondary aerosols but the adjoint method is able to attribute them to precursor emissions. Future research is needed to reconcile the source category definitions among different source attribution methods.

Lastly, when we consider the source influences summed by region, the geographical distribution of adjoint sensitivity indicates strong pollution transport from the southern Hebei Province and from the west (Shanxi Province and Inner Mongolia; see the top left panel of figure 4 for the regions) to Beijing. The transport pathways are consistent with results from backtrajectory models (Wang *et al* 2004, Zhang *et al* 2013). Summing sensitivities over the municipality of Beijing accounts for 52.8% of the total sensitivity. Sources over Hebei Province account for 26.1% of the total sensitivity, 6.7% from Tianjin, and the rest 14% transported from sources further away.

We further apply the adjoint to Tianjin (39.13°N, 117.20°E) and Shijiazhuang (38.05°N, 114.48°E; capital city of Heibei Province) over North China. The left panels of figure 4 show the sensitivities of mean surface PM2.5 concentrations at Beijing, Tianjin, and Shijiazhuang to the spatial distribution of aerosol emissions for January 2013–2015. The right panels of figure 4 summarize the percent contributions integrating geographically over municipalities of Beijing, Tianjin, Hebei and other regions. All three cities show strong regional pollution transport influences spreading over North China. Beijing and Tianjin show local PM2.5 source contributions of 45–53%, and large contributions of ~26% from sources over Hebei. PM2.5 over Shijiazhuang is less impacted by sources in Beijing (less than 2%), but show considerable influences from Shanxi Province (13%) owing to the prevailing southwest winds (Wang *et al* 2014a).

The Action Plan on Air Pollution Prevention and Control released by the Chinese State Council in September 2013 set a strict target for the Beijing–Tianjin– Hebei region with the PM2.5 concentrations to be reduced by 25% by 2017 relative to 2012 (Chinese State Council 2013). We can see for Beijing in January, it would require at least 46% reduction of all emissions in Beijing if only a local control policy is applied, or on average 29% reduction of emissions in the Beijing– Tianjin–Hebei region with a joint regional pollution control policy.

# 4. Conclusions

In summary, we have applied the adjoint method to quantify the sources contributing to surface PM2.5 pollution over North China from January 2013 to 2015 using the GEOS-Chem chemical transport model at  $1/4^{\circ} \times 5/16^{\circ}$  horizontal resolution. We show that wintertime surface PM2.5 over Beijing is largely contributed by emissions of organic carbon, anthropogenic fine dust, and SO<sub>2</sub> mainly from residential and industrial sources, and by emissions of NH3 primarily from agricultural activities. By identifying the most influential source origins at fine geographical resolution, the adjoint sensitivity map can provide valuable information to design effective air quality strategies to mitigate the PM2.5 air pollution. The results presented here do not address uncertainties in the model emission inventory; this can be improved by assimilating observations to optimize aerosol emissions using the adjoint when more observations become available.

#### Acknowledgments

This work was funded by China's National Basic Research Program (2014CB441303), and by the National Natural Science Foundation of China (41205103 and 41475112).

#### References

- Alexander B *et al* 2009 Transition metal-catalyzed oxidation of atmospheric sulfur: global implications for the sulfur budget *J. Geophys. Res.* **114** D02309
- Bey I *et al* 2001 Global modeling of tropospheric chemistry with assimilated meteorology: model description and evaluation *J. Geophys. Res.* **106** 23073–95
- Capps S L *et al* 2012 ANISORROPIA: the adjoint of the aerosol thermodynamic model ISORROPIA *Atmos. Chem. Phys.* **12** 527–43

- Chen D *et al* 2009 Regional CO pollution and export in China simulated by the high-resolution nested-grid GEOS-Chem model *Atmos. Chem. Phys.* **9** 3825–39
- Chinese State Council 2013 Action Plan on Air Pollution Prevention and Control (in Chinese) (http://gov.cn/zwgk/2013-09/12/ content\_2486773.htm) (accessed on 6 May 2014)
- Clifton C L *et al* 1988 Rate constant for the reaction of  $NO_2$  with sulfur (IV) over the pH range 5.3–13 *Environ. Sci. Technol.* 22 586–9
- Fairlie T D, Jacob D J and Park R J 2007 The impact of transpacific transport of mineral dust in the United States Atmos. Environ. 41 1251–66
- Fairlie T D *et al* 2010 Impact of mineral dust on nitrate, sulfate, and ozone in transpacific Asian pollution plumes *Atmos. Chem. Phys.* **10** 3999–4012
- Fountoukis C and Nenes A 2007 ISORROPIA II: a computationally efficient thermodynamic equilibrium model for K<sup>+</sup>-Ca<sup>2+</sup>-Mg<sup>2+</sup>-NH<sub>4</sub><sup>4</sup>-Na+-SO<sub>4</sub><sup>2-</sup>-NO<sub>3</sub><sup>-</sup>-Cl<sup>-</sup>-H<sub>2</sub>O aerosols *Atmos. Chem. Phys.* 7 4639–59
- Fu T-M *et al* 2012 Carbonaceous aerosols in China: top-down constraints on primary sources and estimation of secondary contribution *Atmos. Chem. Phys.* **12** 2725–46
- Gordon T D *et al* 2014 Secondary organic aerosol formation exceeds primary particulate matter emissions for light-duty gasoline vehicles *Atmos. Chem. Phys.* **14**4661–78
- Guo S et al 2014 Elucidating severe urban haze formation in China Proc. Natl Acad. Sci. USA 111 17373–8
- He H et al 2014 Mineral dust and NO<sub>x</sub> promote the conversion of SO<sub>2</sub> to sulfate in heavy pollution days *Sci. Rep.* **4** 4172
- Henze D K et al 2007 Development of the adjoint of GEOS-Chem Atmos. Chem. Phys. 7 2413–2413
- Henze D K *et al* 2009 Inverse modeling and mapping US air quality influences of inorganic PM2.5 precursor emissions with the adjoint of GEOS-Chem *Atmos. Chem. Phys.* **9** 5877–903
- Huang R J *et al* 2014 High secondary aerosol contribution to particulate pollution during haze events in China Nature 514 218–22
- Jacob D J 2000 Heterogeneous chemistry and tropospheric ozone Atmos. Environ. 34 2131–59
- Kharol S K *et al* 2013 Persistent sensitivity of Asian aerosol to emissions of nitrogen oxides *Geophys. Res. Lett.* **40** 1–6
- Kopacz M *et al* 2009 Comparison of adjoint and analytical bayesian inversion methods for constraining Asian sources of carbon monoxide using satellite (MOPITT) measurements of CO columns J. Geophys. Res. 114 D04305
- Lee Y N and Schwartz S E 1983 Kinetics of oxidation of aqueous sulfur(IV) by nitrogen dioxide *Precipitation Scavenging, Dry Deposition and Resuspension* vol 1 ed H R Pruppacher *et al* (New York: Elsevier)
- Lei Y *et al* 2011 Primary anthropogenic aerosol emission trends for China, 1990–2005 *Atmos. Chem. Phys.* 11 931–54
- Littlejohn D et al 1993 Oxidation of aqueous sulfite ion by nitrogen dioxide Environ. Sci. Technol. 27 2162–7
- Martin R V *et al* 2003 Global and regional decreases in tropospheric oxidants from photochemical effects of aerosols *J. Geophys. Res.* **108** D34097
- Nordin E Z *et al* 2013 Secondary organic aerosol formation from idling gasoline passenger vehicle emissions investigated in a smog chamber *Atmos. Chem. Phys.* **13** 6101–16
- Park R J *et al* 2004 Natural and transboundary pollution influences on sulfate-nitrate-ammonium aerosols in the United States: implications for policy J. Geophys. Res. **109** D15204
- Sarwar G *et al* 2013 Potential impacts of two SO<sub>2</sub> oxidation pathways on regional sulfate concentrations: aqueous-phase oxidation by NO<sub>2</sub> and gas-phase oxidation by stabilized criegee intermediates *Atmos. Environ.* **68** 186–97
- Seinfeld J H and Pandis S N 2006 Atmospheric Chemistry and Physics: From Air Pollution to Climate Change (New York: Wiley)
- Tao J et al 2014 PM2.5 pollution in a megacity of southwest China: source apportionment and implication Atmos. Chem. Phys. 14 8679–99
- van Donkelaar A *et al* 2010 Global estimates of ambient fine particulate matter concentrations from satellite-based

aerosol optical depth: development and application *Environ*. *Health Perspect*. **118** 847–55

- Wang J et al 2012 Top-down estimate of dust emissions through integration of MODIS and MISR aerosol retrievals with the GEOS-Chem adjoint model *Geophys. Res. Lett.* **39** L08802
- Wang L T *et al* 2014a The 2013 severe haze over southern Hebei, China: model evaluation, source apportionment, and policy implications *Atmos. Chem. Phys.* **14** 3151–73
- Wang S *et al* 2011 Impact assessment of ammonia emissions on inorganic aerosols in East China using response surface modeling technique *Environ. Sci. Technol.* **45** 9293–300
- Wang Y et al 2013 Sulfate-nitrate-ammonium aerosols over China: response to 2000–2015 emission changes of sulfur dioxide, nitrogen oxides, and ammonia Atmos. Chem. Phys. 13 2635–52
- Wang Y *et al* 2014b Enhanced sulfate formation during China's severe winter haze episode in January 2013 missing from current models *J. Geophys. Res.* **119** 10425–40
- Wang Y Q *et al* 2004 The transport pathways and sources of PM10 pollution in Beijing during spring 2001, 2002 and 2003 *Geophys. Res. Lett.* **31** L14110
- Zhang J K *et al* 2014a Characterization of submicron aerosols during a month of serious pollution in Beijing, 2013 *Atmos. Chem. Phys.* 14 2887–903
- Zhang L et al 2012 Nitrogen deposition to the United States: distribution, sources, and processes Atmos. Chem. Phys. 12 4539–54

- Zhang L *et al* 2014b Sources contributing to background surface ozone in the US intermountain West *Atmos. Chem. Phys.* **14** 5295–309
- Zhang Q et al 2007 Major components of China's anthropogenic primary particulate emissions Environ. Res. Lett. 2045027
- Zhang Q et al 2009 Asian emissions in 2006 for the NASA INTEX-B mission Atmos. Chem. Phys. 9 5131–53
- Zhang R *et al* 2013 Chemical characterization and source apportionment of PM2.5 in Beijing: seasonal perspective *Atmos. Chem. Phys.* **13** 7053–74
- Zhang X Y *et al* 2012 Atmospheric aerosol compositions in China: spatial/temporal variability, chemical signature, regional haze distribution and comparisons with global aerosols *Atmos. Chem. Phys.* **12** 779–99
- Zhao H Y *et al* 2015 Atmospheric nitrogen deposition to the northwestern pacific: seasonal variation and source attribution *Atmos. Chem. Phys. Discuss.* **15** 13657–703
- Zheng B *et al* 2015 Heterogeneous chemistry: a mechanism missing in current models to explain secondary inorganic aerosol formation during the january 2013 haze episode in North China Atmos. Chem. Phys. 15 2031–49
- Zhu L *et al* 2013 Constraining US ammonia emissions using TES remote sensing observations and the GEOS-Chem adjoint model *J. Geophys. Res.* **118** 3355–68