

1 **Numerical analysis of agricultural emissions impacts on PM_{2.5} in China using a high-**
2 **resolution ammonia emission inventory**

3 Xiao Han^{1,2}, Lingyun Zhu⁵, Mingxu Liu⁴, Yu Song⁴, Meigen Zhang^{1,2,3},

4 ¹*State Key Laboratory of Atmospheric Boundary Layer Physics and Atmospheric Chemistry, Institute of*
5 *Atmospheric Physics, Chinese Academy of Sciences, Beijing 100029, China*

6 ²*College of Earth and Planetary Sciences, University of Chinese Academy of Sciences, Beijing 100049,*
7 *China*

8 ³*Center for Excellence in Urban Atmospheric Environment, Institute of Urban Environment, Chinese*
9 *Academy of Sciences, Xiamen 361021, China*

10 ⁴*State Key Joint Laboratory of Environmental Simulation and Pollution Control, Department of*
11 *Environmental Science, Peking University, Beijing 100871, China.*

12 ⁵*Shanxi Province Institute of Meteorological Sciences, Taiyuan 030002, China*

13

14

15

16

17

18

19

20

21

22

23

24

25

26

27

28

29

30 **Abstract**

31 China is one of the largest agricultural countries in the world. Thus, NH₃ emission from agricultural
32 activities in China considerably affects the country's regional air quality and visibility. In this study, a high-
33 resolution agricultural NH₃ emission inventory compiled on 1 km × 1 km horizontal resolution was applied
34 to calculate the NH₃ mass burden in China and reliably estimate the influence of NH₃ on agriculture. The
35 key parameter emission factors of this inventory was enhanced by considering many experiment results,
36 and the dynamic data of spatial and temporal information were updated using statistical data of 2015. In
37 addition to fertilizers and husbandry, farmland ecosystems, livestock waste, crop residue burning, fuel wood
38 combustion, and other NH₃ emission sources were included in this inventory. Furthermore, a source
39 apportionment tool, namely, Integrated Source Apportionment Method (ISAM) coupled with the air quality
40 modeling system Regional Atmospheric Modeling System and Community Multiscale Air Quality, was
41 applied to capture the contribution of NH₃ emitted from total agriculture (Tagr) in China. The aerosol mass
42 concentration in 2015 was simulated, and results showed that the high mass concentration of NH₃ exceeded
43 10 μg m⁻³ and mainly appeared in the North China Plain, Central China, Yangtze River Delta, and Sichuan
44 Basin. Moreover, the annual average contribution of Tagr NH₃ to PM_{2.5} mass burden was 14%–22% in
45 China. Specific to the PM_{2.5} components, Tagr NH₃ contributed dominantly to ammonium formation
46 (87.6%) but trivially to sulfate formation (2.2%). In addition, several brute-force sensitive tests were
47 conducted to estimate the impact of Tagr NH₃ emission reduction on PM_{2.5} mass burden. In contrast to the
48 result of ISAM, even though the Tagr NH₃ only provided 10.1% contribution to nitrate under the current
49 emission scenario, the reduction of nitrate could reach 95.8% upon removal of the Tagr NH₃ emission. This
50 deviation occurred because the contribution of NH₃ to nitrate should be small under a “rich
51 NH₃” environment and large under a “poor NH₃” environment. Thus, the influence of NH₃ on nitrate
52 formation would be enhanced with the decrease in ambient NH₃ mass concentration.

53

54

55

56

57

58

59

1. Introduction

Ammonia (NH_3) is an important pollution species, which is a principal neutralizing agent for acid aerosols SO_4^{2-} and NO_3^- that are formed from SO_2 and NO_x (Chang, 1989; McMurry et al., 1983). In addition, NH_3 influences the rate of particle nucleation (Ball et al., 1999; Kulmala et al., 2002) and enhances secondary organic aerosol (SOA) yield (Babar et al., 2017). Widespread haze events have frequently occurred in most regions of eastern China in recent years, and several studies have reported that secondary inorganic salts, including sulfate, nitrate, and ammonium, form the majority of total aerosols in the urban and rural regions (Tao et al., 2014; Wang et al., 2016; Zhang et al., 2012; Lai et al., 2016; Zhang et al., 2018). Therefore, in addition to the heavy emissions of SO_2 and NO_2 , NH_3 emissions from agricultural activities are non-negligible.

China is one of the largest agricultural countries in the world. Even though a decrease appeared from 2006 to 2012, the annual NH_3 emission budget, which reached 9.7–12 Tg (Kang et al., 2016; Xu et al., 2016; Zhou et al., 2015), remains huge and leads to high NH_3 ambient concentration. This massive NH_3 emission considerably affects regional air quality and horizontal visibility. First, the major $\text{PM}_{2.5}$ components, $(\text{NH}_4)_2\text{SO}_4$, $(\text{NH}_4)_3\text{H}(\text{SO}_4)_2$, NH_4HSO_4 , and NH_4NO_3 , were partially or fully yielded from neutralizing H_2SO_4 and HNO_3 via NH_3 reaction (Tanner et al., 1981; Brost et al., 1988; Quan et al., 2014; Zhao et al., 2013; Zhang et al., 2014). Studies also showed that NH_3 improves H_2SO_4 nucleation by 1–10 times (Benson et al., 2011) and provides enough new particles to alter the number and size distributions. Thus, NH_3 and its secondary product NH_4^+ play an important role in the formation of air pollution and haze days. Some research has shown that approximately 80% of total anthropogenic NH_3 emissions is derived from agricultural sources and that livestock manure provides more contributions than that of synthetic fertilizers (Kang et al. 2016; Zhou et al., 2016). The Chinese government has taken several control strategies to reduce particle pollution and its precursors; some examples of these systems include catalytic reduction systems in the power sector (Xia et al., 2016) and measures to change coal to gas for residents' life and heating (Ren et al., 2014). Related observations have shown that the mass burden of SO_2 and NO_x have distinctly decreased in recent years (De Foy et al., 2016; Wang et al., 2015; Zheng et al., 2018). However, no specific measure for the control of agricultural NH_3 emission has yet to be implemented, and the total agricultural NH_3 emission budget was not considerably changed from 2010 to 2017 (Zheng et al., 2018).

In addition, an accurate information of agriculture NH_3 emission is important for estimating the NH_3

90 mass burden and its environmental effect. Several studies have focused on NH₃ emissions from agricultural
91 activities in China or East Asia. The second version of the Regional Emission Inventory in Asia (REAS)
92 has established an anthropogenic emission inventory, which includes the source of agricultural NH₃
93 (fertilizer application and livestock) (Kurokawa et al., 2013). This inventory, which targeted years 2000–
94 2008, has a 0.25×0.25-degree spatial resolution with monthly variation. MASAGE_NH₃ (Magnitude and
95 Seasonality of Agricultural Emissions model for NH₃) developed a bottom-up NH₃ emission inventory by
96 using the adjoint of the GEOS-Chem chemical transport model (Paulot et al., 2014). The network data for
97 NH₄⁺ wet deposition fluxes from 2005 to 2008 were inversed to optimize China's NH₃ emission in this
98 inventory. Fu et al. (2015) used the Community Multiscale Air Quality (CMAQ) model coupled with an
99 agroecosystem to estimate the NH₃ emissions with high spatial and temporal resolution in 2011; the model
100 could obtain hourly emission features through online model calculations. These NH₃ emission inventories
101 have provided useful datasets for understanding the distribution features of NH₃ mass burden in China.
102 However, with the migration of population, economic growth, and the increase in the consumption of
103 agricultural products, the spatial distribution and strength of agricultural NH₃ emission remarkably changed
104 in China during the last decade (Xu et al., 2017); thus, a reliable emission information based on the recent
105 year is also necessary for estimating the NH₃ mass burden.

106 Previous studies have investigated the influence of NH₃ emission to aerosol loading in several areas
107 of China. Wu et al. (2008) conducted sensitivity studies to assess the impact of livestock NH₃ emissions on
108 the PM_{2.5} mass concentration in North China by using MM5/CMAQ modeling system. The results showed
109 that livestock NH₃ provides >20% contribution to nitrate and ammonium but provides minimal contribution
110 to sulfate. Wang et al. (2011) used the response surface modeling technique to estimate the contribution of
111 NH₃ emission in East China and found that the total NH₃ emission contributes 8%–11% to PM_{2.5}
112 concentration and the nonlinear effects are significant while the transition between NH₃ rich and poor
113 conditions. Fu et al. (2017) and Zhao et al. (2017) also investigated the impact of NH₃ emission on PM_{2.5}
114 in East China and Hai River Basin. However, related research remains scarce and mainly focused on the
115 local regions, and most of them generally use the brute-force sensitivity method to estimate the NH₃ impact
116 on the basis of the chemistry model, which reflects the particle concentration change with emission
117 reduction (Koo et al., 2009).

118 A comprehensive high-resolution NH₃ emission inventory PKU-NH₃, which is based on the year 2015,
119 is applied in this study to capture the agricultural NH₃ mass concentration in China. In addition, the

120 contribution to PM_{2.5} particle is estimated via the air quality modeling system Regional Atmospheric
121 Modeling System (RAMS) –CMAQ coupled with the online source tagged module Integrated Source
122 Apportionment Method (ISAM). Compared with previous studies, this high-resolution agricultural NH₃
123 emission inventory is more accurate and reflects the latest spatial and temporal distribution features (Liu et
124 al., 2019). Major trace gases and aerosol species in 2015 are simulated via the modeling system and
125 evaluated using substantial observation data. The contribution to pollutant concentrations can be tagged
126 and quantified by RAMS–CMAQ–ISAM under the current scenario (Wang et al., 2009). Then, several
127 brute-force sensitivity tests are conducted to estimate the effect of reducing agricultural NH₃ emission on
128 the PM_{2.5} mass burden. The results from the source apportionment simulation and brute-force sensitivity
129 tests in January, April, July, and October are presented, and the detailed features of seven major populated
130 areas (as shown in Figure 1) of China are discussed.

132 2. Methodology

133 The emission inventory is described as follows: First, the NH₃ emission data in China were provided
134 by the PKU-NH₃ emission inventory (Kang et al., 2016; Zhang et al., 2018). This inventory was developed
135 on the basis of previous studies (Huang et al., 2012) and improved horizontal resolution and accuracy. It
136 was compiled on 1 km×1 km horizontal resolution with monthly based statistic data in 2015. One of the
137 most uncertain parameters of the emission factors applied in this inventory was enhanced by considering
138 as many native experiment results as possible with ambient temperature, soil acidity, and other factors
139 change. In addition, this inventory not only includes the fertilizer and husbandry emissions from agricultural
140 activities but also collects the emission data of farmland ecosystems, livestock waste, biomass burning
141 (forest and grassland fires, crop residue burning, and fuel wood combustion), and other sources (excrement
142 waste from rural populations, the chemical industry, waste disposal, NH₃ escape from thermal power plants,
143 and traffic sources). Second, the anthropogenic emission of primary aerosols and the precursors were
144 obtained from the MIX Asian emission inventory (base year 2012) prepared by the Model Inter-Comparison
145 Study for Asia (MICS-ASIA III) (Lu et al., 2011; Lei et al., 2011). The anthropogenic emission sources of
146 SO₂, NO_x, volatile organic compounds (VOCs), black carbon, organic carbon, primary PM_{2.5}, and PM₁₀
147 were obtained from the monthly-based MIX inventory with 0.25°× 0.25° spatial resolution. The REAS
148 version 2 (Kurokawa et al., 2013) and Global Fire Emissions Database version 3 (van der Werf et al., 2010)
149 were used to provide data on VOCs, nitrogen oxides from flight exhaust, lighting, paint, wildfires, savanna

150 burning, and slash-and-burn agriculture.

151 The modeling system RAMS–CMAQ was applied to simulate the transformation and transport of
152 pollutants in the atmosphere. The CMAQ (version 5.0.2) released by the US Environmental Protection
153 Agency (Eder et al., 2009; Mathur et al., 2008) was the major component of the RAMS–CMAQ modeling
154 system. In this model, the CB05 (version CB05tucl) chemical mechanism (Whitten, 2010) was used to treat
155 the gas-phase chemical mechanism. The simulation of O₃ in urban plumes, which could impact the NO_x
156 chemical transformation and fine particle mass predictions, was updated in this version to obtain reasonable
157 results. The sixth-generation model CMAQ aerosol model (AERO6), which added nine new PM_{2.5} species
158 and updated the SOA yield parametrization and primary organic aerosol aging processes, was used to
159 simulate the formation and dynamic processes of aerosols. The ISORROPIA model (version 2.1)
160 (Fountoukis and Nenes, 2007) was used to describe the thermodynamic equilibrium of gas-particle
161 transformation. The highly versatile numerical model RAMS, which can well capture the boundary layer
162 and the underlying surface, was applied to provide the meteorological fields for CMAQ (Cotton et al.,
163 2003). The European Centre for Medium-Range Weather Forecasts reanalysis datasets (1°×1° spatial
164 resolution) were used to supply the background fields and sea surface temperatures. The model domain
165 (Figure 1) is 6654 km×5440 km with 64 km² fixed grid cells and uses a rotated polar stereographic map
166 projection, which covers the entire mainland of China and its surrounding regions. The model has 15
167 vertical layers, and half of them are located in the lowest 2 km to provide a precise simulation of the
168 atmospheric boundary layer.

169 The ISAM is a flexible and efficient online source apportionment implementation that was used to
170 track multiple pollutants emitted from different geographic regions and source types. Compared with its
171 previous version (i.e., tagged species source apportionment), the processes of tracking tagged tracer
172 transport and precursor reaction were optimized to balance the computational requirements and reliable
173 representation of physical and chemical evolution. To reduce the nonlinear effect during phase
174 transformation and relative chemical interactions, a standalone subroutine “wrapper” approach was applied
175 in ISAM to apportion the secondary PM species and their precursor gases during the thermodynamic
176 equilibrium simulation; a hybrid approach, which uses LU decomposition triangular matrices (Yang et al.,
177 1997), was developed for describing gas-phase chemical interactions. In this study, ISAM was coupled into
178 RAMS–CMAQ and set to trace the transport and chemical reactions of NH₃ from fertilizer and husbandry
179 emission sectors and quantitatively estimate the contribution of agriculture NH₃ emission to the PM_{2.5} mass

180 concentration in China.

182 **3. Model evaluation**

183 To evaluate the performance of the model, substantial observation data are used for comparison with
184 the simulation results. Meteorological factors are important to capture the formation processes and transport
185 of secondary aerosols. Thus, in this study, the observed meteorological data from surface stations of the
186 Chinese National Meteorological Center were collected to evaluate the performance of the model. Detailed
187 information is provided in Appendix A. Furthermore, the observed SO₂, NO₂, and PM_{2.5} released from the
188 Ministry of Environmental Protection of China were applied to evaluate the modeled mass concentration
189 of these pollutants. The hourly observation data in January, April, July, and October at six stations located
190 in Beijing, Jinan, Shijiazhuang, Nanjing, Guangzhou, and Zhengzhou were collected in this study. The
191 scatter plots of comparison are shown in Figure 2, and the statistical parameters between the observations
192 and simulations are listed in Tables 1–3. Most of the scatter points broadly gather around the 1:1 solid line.
193 Most of the correlation coefficients in Table 1–3 are higher than 0.5, indicating that the model can capture
194 the regional variation features of measurements. The standard deviations between the observed and
195 simulated results are similar in most cases as well. The simulation results performed better in winter
196 compared with that in summer because the diffusion condition was strong and the mass concentration
197 changed noticeably during summer time. The modeled PM_{2.5} generally performed well due to relatively
198 high correlation coefficients. The evident deviation of the modeled mean, which was higher than that of the
199 observation, was between the observed and modeled SO₂. The emission of SO₂ reduced rapidly because of
200 the control measures from 2013 in China. However, the emission inventory may not reflect this feature and
201 may slightly overestimate the mass burden.

202 The horizontal distributions of the modeled monthly NH₃ mass concentration in January, April, July,
203 and October in 2015 are shown in Figure 3. Pan et al. (2018) provided the distributions of satellite NH₃
204 total column distribution and the surface NH₃ concentrations at several observation sites in Figure 1 of their
205 paper. Their results showed that the highest mass burden is concentrated in the North China Plain (NCP),
206 Central China (CNC), Yangtze River Delta (YRD), and Sichuan Basin (SCB). The simulation results in this
207 study broadly reflect these distribution features. The values of NH₃ concentrations in these regions could
208 reach 10–25 μg m³ in Pan et al. (2018); these results coincided well with the simulation results. However,
209 some considerable deviation appeared in areas of the eastern part of Gansu Province. In this study, the

210 modeled NH₃ in these regions was slightly higher than those of the observations in Pan et al. (2018). Zhang
211 et al. (2018) also presented the NH₃ mass concentration in four seasons over China through simulation
212 (horizontal distribution) and ground-based measurements (point values) in Figure 9 of their paper. In
213 addition to the regions maintained in Pan et al. (2018), the high mass burden of NH₃ also appeared in
214 Northeast China (NEC), as shown by the simulation and observation results in Zhang et al. (2018).
215 Generally, this distribution feature should be reasonable because the Three River Plain located in NEC is
216 an important agricultural base in China, and the NH₃ emission in this region can be strong during spring
217 and summer. The simulation results in this study also followed the seasonal variation feature of NH₃ mass
218 burden, as shown in Zhang et al. (2018); the feature was higher in summer and lower in winter, and the
219 magnitude was also close with each other. Thus, the modeled NH₃ concentration measured by RAMS–
220 CMAQ is reliable and can be applied for the analysis in this study.

221

222 **4. Results and discussions**

223 The horizontal distributions of modeled monthly PM_{2.5} mass concentrations in January, April, July,
224 and October in 2015 and the surface wind field are shown in Figure 4. Over the eastern part of China, the
225 heavy PM_{2.5} pollution happened in January, and the relatively better air quality appeared in July. The large
226 PM_{2.5} mass burden exceeded 200 μg m⁻³ in January and was mainly concentrated in the NCP, the Yangtze
227 River Valley of CNC, and SCB; these observations broadly coincided with the regions covered by a high
228 mass burden of NH₃, as shown in Figure 3. The wind speed in the regions mentioned above was relatively
229 weak, implying that the diffusion condition was poor, and more aerosols can be trapped in these regions. In
230 addition, the PM_{2.5} mass burden (50–150 μg m⁻³) in July was lower than that of other months. Considering
231 that NH₃ emission is mainly concerned with secondary inorganic aerosols (SNA), such as sulfate, nitrate,
232 and ammonium formation, the analysis hereafter mainly focuses on SNA. Figure 5 presents the modeled
233 monthly SNA mass concentration in January, April, July, and October in 2015. The mass loading of SNA
234 generally provided 40%–60% to the total PM_{2.5} in the eastern part of China; this result is comparable with
235 previous studies (Cao et al., 2017; Chen et al., 2016; Lai et al., 2016; Wang et al., 2016). The distribution
236 pattern and seasonal variation of SNA also followed the features of PM_{2.5}, and the high mass concentration
237 of SNA could exceed 100 μg m⁻³ in January.

238 Then, the contributions of NH₃ from multiple agricultural emissions (including fertilizer, husbandry,
239 farmland ecosystems, livestock waste, crop residue burning, and excrement waste from rural populations)

240 to aerosols were calculated using RAMS–CMAQ–ISAM; the monthly average contribution percentage of
241 total agricultural activities (Tagr) in January, April, July, and October is shown in Figure 6. Generally, the
242 Tagr NH₃ provided 30%–50% contribution in January and October and 20%–40% contribution in April and
243 July to SNA over most parts of eastern China. The relatively lower value mainly appeared in April.

244 The regional average percentage of Tagr contribution to sulfate, nitrate, ammonium, SNA, and PM_{2.5}
245 are shown in Table 4. As shown in this table, the annual average Tagr NH₃ provided major contributions,
246 which reached approximately 90%, to ammonium and relatively small contribution (5%–10%) to nitrate
247 mass burden. However, the contribution to sulfate was minimal because sulfate formation from SO₂ can
248 occur in various ways, in addition to neutralization by NH₃, such as oxidized by H₂O₂, O₃, or peroxyacetic
249 acid. The seasonal variation of ammonium was evident; it could be higher than 99% in January but lower
250 than 70% in July. Most of the differences as shown in Table 4 could exceed 10% because the NH₃ emitted
251 from other sources (anthropogenic and natural sources) was substantial in these regions during summer.
252 The annual average Tagr NH₃ provided 20%–40% contribution to SNA mass concentration, and the
253 contributions in January were larger than that in July. The seasonal variation and spatial features of Tagr
254 NH₃ contribution to PM_{2.5} mass concentration were similar with the features of SNA and generally provided
255 approximately 14%–22% contribution to the total PM_{2.5} mass concentration in these places. By contrast,
256 the annual contribution in China was higher than those in the regions mentioned above. This feature
257 indicates that the Tagr NH₃ provided more contribution compared with other sources over regions with
258 weaker anthropogenic activities.

259 In addition, the brute-force method (zero-out sensitivity test), which can capture the effect of emission
260 change on aerosol mass burden, was applied to investigate the impact of the removal of Tagr NH₃ emission.
261 In contrast to online source apportionment, the brute-force method mainly reflects the disparity of chemical
262 balance caused by the emission change, which could considerably alter secondary pollutant formation.
263 Several sensitivity tests were conducted, and the results are shown in Figure 7 and Table 5. Figure 7 presents
264 the mass burden variation of SNA associated with the Tagr NH₃ removal. Figure 7 shows that the reduction
265 pattern and seasonal variation of the aerosol were broadly followed by those of their mass burden. The
266 considerable reduction of SNA mainly appeared in the high-concentration regions and generally exceeded
267 25 μg m⁻³. Table 5 shows the percentage of the variation of sulfate, nitrate, ammonium, SNA, and PM_{2.5}.
268 Compared with Table 4, the variation percentage of SNA and PM_{2.5} reached 30%–60% and 24%–42%,
269 respectively, and approximately two times higher than those of the contribution percentage. This

270 remarkable distinction was mainly caused by the variation of nitrate; that is, the contribution of Tagr NH₃
271 to nitrate was generally below 10%, as shown in Table 4, but the reduction of nitrate associated with
272 removing Tagr NH₃ emission could exceed 90%, as shown in Table 5. This difference between the results
273 of ISAM and brute-force tests was expected due to the high nonlinearity in the NO_x chemistry. The nitrate
274 formation could become more sensitive when the “rich NH₃” environment shifts to a “poor NH₃”
275 environment, which means the decrease in nitrate mass burden would accelerate with NH₃ emission
276 reduction. Therefore, it can be deduced that the contribution of NH₃ to nitrate should be remarkably lower
277 under “rich NH₃” environments compared with that under “poor NH₃” environments. A similar
278 phenomenon was also reported by some previous studies (Wang et al., 2011; Xu et al., 2016). To prove this
279 point, more brute-force sensitivity tests were conducted. The variation of sulfate, nitrate, ammonium, and
280 SNA mass burden associated with the reduction of NH₃ emission (80%, 50%, 40%, 30%, 20%, and 10%
281 Tagr NH₃ emission, respectively) is shown in Figure 8. The decline of nitrate mass concentration was more
282 rapid than that of ammonium, and the trend became slightly faster with the reduction of NH₃ emission
283 (signified from “rich NH₃” to “poor NH₃” environments) in most regions. The acceleration of nitrate mass
284 burden decline was substantial in regions with strong NH₃ emission. Furthermore, this acceleration stopped
285 when 20% of NH₃ emission remained, as shown in Figure 8.

286

287 **5. Conclusions**

288 The emission budget of agricultural NH₃ was huge and played an important role on the regional
289 particle pollution in China. As a precursor of the secondary aerosol, the reasonable estimation of the
290 nonlinear processes of secondary aerosol formation should be the key point for capturing the contribution
291 of NH₃ to particle pollution. In this study, the air quality modeling system RAMS-CMAQ was applied to
292 simulate the spatial–temporal distribution of trace gas and aerosols in 2015. In addition, the PKU-NH₃
293 emission inventory, which was compiled on 1 km × 1 km horizontal resolution with monthly based data,
294 was applied to capture the features of agricultural NH₃ emission in China accurately. Then, the source
295 apportionment module ISAM was coupled into this modeling system to estimate the contribution of
296 agricultural NH₃ to PM_{2.5} mass burden quantitatively. Brute-force sensitivity tests were also conducted to
297 discuss the impact of agricultural NH₃ emission reduction. The meteorological factors and mass
298 concentration of NH₃, SO₂, NO₂, and PM_{2.5} from simulation were evaluated and showed consistency with
299 the observation data. Some interesting results were explored and summarized as follows:

300 (1) The high mass burden of NH_3 could exceed $10 \mu\text{g m}^{-3}$ and mainly appeared in the NCP, CNC,
301 YRD, and SCB. These regions had high coincidence with regions that are heavily covered with particle
302 pollution. Therefore, it can be deduced that the influence of agricultural NH_3 on the $\text{PM}_{2.5}$ mass
303 concentration is crucial.

304 (2) The results from ISAM simulation showed that the Tagr NH_3 provided 17%–23% and 15%–22%
305 contribution to the $\text{PM}_{2.5}$ in January and July, respectively, in most parts of eastern China, and the largest
306 annual average contribution appeared in CNC (17.5%). Specific to SNA components, the annual and
307 regional average contributions of Tagr NH_3 to ammonium, nitrate, and sulfate in China were 87.6%, 10.1%,
308 and 2.2%, respectively. Therefore, agricultural NH_3 emission contributes considerably to ammonium
309 formation but minimally to sulfate due to various ways of sulfate formation.

310 (3) The brute-force sensitive test could reflect the effect of changing Tagr NH_3 emission on $\text{PM}_{2.5}$ mass
311 burden. The results indicated that the reduction percentage of $\text{PM}_{2.5}$ mass burden due to removal Tagr NH_3
312 emission could reach 24%–42% in most parts of eastern China; these values are approximately two times
313 higher than the contribution. The nitrate reduction percentage that exceeded 90% was the major reason for
314 this remarkable difference. In addition, further analysis proved that the ambient NH_3 mass burden could
315 affect its contribution to SNA formation, that is, the NH_3 contribution to nitrate should be low under “rich
316 NH_3 ” environments and high under “poor NH_3 ” environments. Therefore, the influence of NH_3 would be
317 enhanced with the decrease in ambient NH_3 mass concentration.

318 This study suggests that the NH_3 influence on the $\text{PM}_{2.5}$ mass burden are complex because of the
319 nonlinearity of secondary aerosol formation. Substantial deviation exists between the results of the ISAM
320 and brute-force methods; thus, these two kinds of results should be distinguished and applied to explain
321 different issues: the contribution under the current scenario and the effect due to emission reduction. The
322 modeling system is a versatile tool that allows us to investigate valuable information for choosing efficient
323 strategies of reducing the impact of agricultural NH_3 and improving air quality.

324 **Acknowledgments**

325 This work was supported by the National Key R&D Programs of China (2017YFC0209803), the
326 National Natural Science Foundation of China (41830109), and the Strategic Priority Research Program of
327 the Chinese Academy of Sciences (XDA19040204).
328
329

Appendix A

The daily average temperature, relative humidity, wind speed and maximum wind direction in January, April, July and October 2015 were compared with the surface shared data from the Chinese National Meteorological Center (<http://data.cma.cn/>) in 9 stations. The comparison results are shown in Figure A1-A4. These stations are located in the East China where the high NH₃ emission regions. Generally, the modeled temperature was in good agreement with the observed data, and can reflect the large fluctuation and seasonal variation of relative humidity as well, except that some of the extreme high or low values appeared abruptly. As shown in Figure A3, most of the daily average wind speed was lower than 3 m s⁻¹ at Zhengzhou, Miyun, Tianjin and Baoding station (all located in the North China Plain), which means the diffusion condition was not good due to the stable weather. Otherwise, the relatively strong wind appeared at Nanjing, Chaoyang, Nanning and Jinan. The modeled wind speed generally reproduced all these features. The direct comparison between observed and modeled wind direction which can be easily influenced by the surrounding surface features is difficult. Nevertheless, the prevailing wind direction in different seasons can be captured by the simulation results for all stations.

In addition, Figure A5 present the regional average NH₃ emission flux (g/s/grid) of different sectors, including fertilizer, Husbandry, Biomass burning, Farmland ecosystems, Waste disposal, and other sectors, over each regions in January, April, July and October. Furthermore, the percent (%) of each NH₃ emission sector was shown in Figure A6. All the information was obtained from the PKU-NH₃ emission inventory directly. It can be seen that the emission flux was higher in summer and lower in winter. The strongest emission flux mainly appeared in BTH, SDP and CNC. The distribution pattern of NH₃ mass concentration

These features generally followed the distribution pattern of NH₃ mass concentration as shown in Figure 3. On the other hand, the major proportion was provided by husbandry and fertilizer, and relatively higher in spring and summer.

Reference

- 360
361 Babar, Z. B., Park, J., and Lim, H.: Influence of NH₃ on secondary organic aerosols from the ozonolysis and photooxidation
362 of a α -pinene in a flow reactor, *Atmos. Environ.*, 164, 71-84, DOI: 10.1016/j.atmosenv.2017.05.034, 2017.
- 363 Ball, S. M., Hanson, D. R., and Eisele, F. L.: McMurry, P. H. Laboratory studies of particle nucleation: Initial results for
364 H₂SO₄, H₂O, and NH₃ vapors, *J. Geophys. Res.*, 104, 23709-23718, DOI: 10.1029/1999JD900411, 1999.
- 365 Benson, D. R., Yu, J. H., Markovich, A., and Lee, S. H.: Ternary homogeneous nucleation of H₂SO₄, NH₃, and H₂O under
366 conditions relevant to the lower troposphere, *Atmos. Chem. Phys.*, 11, 4755-4766, DOI: 10.5194/acp-11-4755-2011,
367 2011.
- 368 Brost, R. A., Delany, A. C., and Huebert, B. J.: Numerical modeling of concentrations and fluxes of HNO₃, NH₃, and
369 NH₄NO₃ near the ground, *J. Geophys. Res.*, 93, 7137-7152, DOI: 10.1029/JD093iD06p07137, 1988.
- 370 Cao, Z., Zhou, X., Ma, Y., Wang, L., Wu, R., Chen, B., and Wang, W.: The concentrations, formations, relationships and
371 modeling of sulfate, nitrate and ammonium (SNA) aerosols over China, *Aerosol Air Qual. Res.*, 17, 84-97, DOI:
372 10.4209/aaqr.2016.01.0020, 2017.
- 373 Chen, Y., Schleicher, N., Cen, K., Liu, X., Yu, Y., Zibat, V., Dietze, V., Fricker, M., Kaminski, U., Chen, Y., Chai, F., and
374 Norra, S.: Evaluation of impact factors on PM_{2.5} based on long-term chemical components analyses in the megacity
375 Beijing, China, *Chemosphere*, 155, 234-242, DOI: 10.1016/j.chemosphere.2016.04.052, 2016.
- 376 Chang, J.: The role of H₂O and NH₃ on the formation of NH₄NO₃ aerosol particles and De-NO_x under the corona discharge
377 treatment of combustion flue gases, *J. Aerosol Sci.*, 20, 1087-1090, DOI: 10.1016/0021-8502(89)90768-4, 1989.
- 378 Cotton, W., Pielke, R., Walko, G., Liston, G., Tremback, C., Jiang, H., McAnelly, R., Harrington, J., Nicholls, M., Carrio,
379 G., and McFadden, J.: RAMS 2001: current status and future directions, *Meteorol. Atmos. Phys.*, 82, 5-29, DOI:
380 10.1007/s00703-001-0584-9, 2003.
- 381 DeFoy, B., Lu, Z., and Streets, D. G.: Satellite NO₂ retrievals suggest China has exceeded its NO_x reduction goals from
382 the twelfth five-year plan, *Sci. Rep.*, 6, 35912, DOI: 10.1007/s00703-001-0584-9, 2016.
- 383 Eder, B., and Yu S.: A performance evaluation of the 2004 release of Models-3 CMAQ, *Atmos. Environ.*, 40, 4811-4824,
384 DOI: 10.1016/j.atmosenv.2005.08.045, 2006.
- 385 Fountoukis, C., and Nenes, A.: ISORROPIA II: a computationally efficient thermodynamic equilibrium model for K⁺-
386 Ca²⁺-Mg²⁺-NH₄⁺-Na⁺-SO₄²⁻-NO₃⁻-Cl-H₂O aerosols, *Atmos. Chem. Phys.*, 7, 4639-4659, DOI: 10.5194/acp-7-4639-
387 2007, 2007.
- 388 Fu, X., Wang, S., Xing, J., Zhang, X., Wang, T., and Hao, J.: Increasing Ammonia Concentrations Reduce the Effectiveness
389 of Particle Pollution Control Achieved via SO₂ and NO_x Emissions Reduction in East China, *Environ. Sci. Technol.*,
390 4, 221-227, DOI: 10.1021/acs.estlett.7b00143, 2017.
- 391 Huang, X., Song, Y., Li, J., Huo, Q., Cai, X., Zhu, T., Hu, M., and Zhang, H.: A high-resolution ammonia emission
392 inventory in China, *Global Biogeochem. Cy.*, 26, 1030-1044, DOI: 10.1029/2011GB004161, 2012.
- 393 Kang, Y., Liu, M., Song, Y., Huang, X., Yao, H., Cai, X., Zhang, H., Kang, L., Liu, X., Yan, X., He, H., Zhang, Q., Shao,
394 M., and Zhu, T.: High-resolution ammonia emissions inventories in China from 1980 to 2012, *Atmos. Chem. Phys.*,
395 16, 2043-2058, DOI: 10.5194/acpd-15-26959-2015, 2016.
- 396 Koo, B., Wilson, G., Morris, R., Dunker, A., and Yarwood, G.: Comparison of Source Apportionment and Sensitivity
397 Analysis in a Particulate Matter Air Quality Model, *Environ. Sci. Technol.*, 43, 6669-6675, 2009.
- 398 Kurokawa, J., Ohara, T., Morikawa, T., Hanayama, S., Maenhout, G., Fukui, T., Kawashima, K., and Akimoto, H.:
399 Emissions of air pollutants and greenhouse gases over Asian regions during 2000-2008: Regional Emission inventory
400 in ASia (REAS) version 2, *Atmos. Chem. Phys.*, 13, 11019-11058, DOI: 10.5194/acp-13-11019-2013, 2013.
- 401 Kulmala, M., Korhonen, P., Napari, I., Karlsson, A., Berresheim, H., and O'Dowd, C. D.: Aerosol formation during
402 PARFORCE: Ternary nucleation of H₂SO₄, NH₃, and H₂O, *J. Geophys. Res.*, 107, DOI: 10.1029/2001JD000900,
403 2002.

404 Lai, S., Zhao, Y., Ding, A., Zhang, Y., Song, T., Zheng, J., Ho, K. F., Lee, S., and Zhong, L.: Characterization of PM_{2.5} and
405 the major chemical components during a 1-year campaign in rural Guangzhou, Southern China, *Atmos. Res.*, 167,
406 208-215, DOI: 10.1016/j.atmosres.2015.08.007, 2016.

407 Lei, Y., Zhang, Q., He, K., and Streets, D.: Primary anthropogenic aerosol emission trends for China, 1990-2005, *Atmos.*
408 *Chem. Phys.*, 11, 931-954, DOI: 10.5194/acp-11-931-2011, 2011.

409 Liu, M., Huang, X., Song, Y., Tang, J., Cao, J., Zhang, X., Zhang, Q., Wang, S., Xu, T., Kang, L., Cai, X., Zhang, H., Yang,
410 F., Wang, H., Yu, J., Lau, A., He, L., Huang, X., Duan, L., Ding, A., Xue, L., Gao, J., Liu, B., and Zhu, T.: Ammonia
411 emission control in China would mitigate haze pollution and nitrogen deposition, but worsen acid rain, *PNAS*, 116,
412 7760-7765, DOI: 10.1073/pnas.1814880116, 2019.

413 Lu, Z., Zhang, Q., and Streets, D. G., Sulfur dioxide and primary carbonaceous aerosol emissions in China and India,
414 1996-2010, *Atmos. Chem. Phys.*, 11, 9839-9864, DOI: 10.5194/acp-11-9839-2011, 2011.

415 Mathur, R., Yu, S., Kang, D., and Schere, K.: Assessment of the winter-time performance of developmental particulate
416 matter forecasts with the Eta-CMAQ modeling system, *J. Geophys. Res.*, 113, DOI: 10.1029/2007JD008580, 2008.

417 McMurry, P. H., Takano, H., and Anderson, G. R.: Study of the ammonia (gas)-sulfuric acid (aerosol) reaction rate, *Environ.*
418 *Sci. Technol.*, 17, 347-352, DOI: 10.1021/es00112a008, 1983.

419 Paulot, F., Jacob, D. J., Pinder, R. W., Bash, J. O., Travis, K., and Henze, D. K.: Ammonia emissions in the United States,
420 European Union, and China derived by high-resolution inversion of ammonium wet deposition data: Interpretation
421 with a new agricultural emissions inventory (MASAGE_NH3), *J. Geophys. Res.*, 119, 4343-4364, DOI:
422 10.1002/2013JD021130, 2014.

423 Pen, Y., Tian, S., Zhao, Y., Zhang, L., Zhu, X., Gao, J., Huang, W., Zhou, Y., Song, Y., Zhang, Q., and Wang, Y.: Identifying
424 ammonia hotspots in China using a national observation network, *Environ. Sci. Technol.*, doi:
425 10.1021/acs.est.7b05235, DOI: 10.1021/acs.est.7b05235, 2008.

426 Quan, J., Tie, X., Zhang, Q., Liu, Q., Li, X., Gao, Y., and Zhao, D.: Characteristics of heavy aerosol pollution during the
427 2012–2013 winter in Beijing, China, *Atmos. Environ.*, 88, 83-89, DOI: 10.1016/j.atmosenv.2014.01.058, 2014.

428 Ren, H., Zhang, L., and Hong, X.: Politic recommendations on strengthening reduction of air pollutant emissions in China,
429 *Environ. Sustain. Dev.*, 39, 4-13, 2014 (in Chinese).

430 Tao, M., Chen, L., Xiong, X., Zhang, M., Ma, P., Tao, J., and Wang, Z.: Formation process of the widespread extreme haze
431 pollution over northern China in January 2013: Implications for regional air quality and climate, *Atmos. Environ.*, 98,
432 417-425, DOI: 10.1016/j.atmosenv.2014.09.026, 2014.

433 Tanner, R. L., Leaderer, B. P., and Spengler, J. D., Acidity of atmospheric aerosols, *Environ. Sci. Technol.*, 15, 1150–1153,
434 DOI: 10.1021/es00092a003, 1981.

435 van der Werf, G., Randerson, J. Giglio, L., Collatz, G., Mu, M., Kasibhatla, P., Morton, D., Defries, R., Jin, Y., and van
436 Leeuwen, T.: Global fire emissions and the contribution of deforestation, savanna, forest, agricultural, and peat fires
437 (1997-2009), *Atmos. Chem. Physics.*, 10, 11707-11735, DOI: 10.5194/acp-10-11707-2010, 2010.

438 Wang, G., Zhang, R., Gomez, M. E., Yang, L., Levy Zamora, M., Hu, M., Lin, Y., Peng, J., Guo, S., Meng, J., Li, J., Cheng,
439 C., Hu, T., Ren, Y., Wang, Y., Gao, J., Cao, J., An, Z., Zhou, W., Li, G., Wang, J., Tian, P., Marrero-Ortiz, W., Secret,
440 J., Du, Z., Zheng, J., Shang, D., Zeng, L., Shao, M., Wang, W., Huang, Y., Wang, Y., Zhu, Y., Li, Y., Hu, J., Pan, B.,
441 Cai, L., Cheng, Y., Ji, Y., Zhang, F., Rosenfeld, D., Liss, P. S., Duce, R. A., Kolb, C. E., and Molina, M. J.: Persistent
442 sulfate formation from London Fog to Chinese haze, *Proc. Natl. Acad. Sci.*, 113, 13630-13635, DOI:
443 10.1073/pnas.1616540113, 2016.

444 Wang, H., Qiao, L., Lou, S., Zhou, M., Ding, A., Huang, H., Chen, J., Wang, Q., Tao, S., Chen, C., Li, L., and Huang, C.:
445 Chemical composition of PM_{2.5} and meteorological impact among three years in urban Shanghai, China, *J. Clean.*
446 *Prod.*, 112, 1302-1311, DOI: 10.1016/j.jclepro.2015.04.099, 2016.

447 Wang, S., Zhang, Q., Martin, R.V., Philip, S., Liu, F., Li, M., Jiang, X., and He, K.: Satellite measurements oversee China's

448 sulfur dioxide emission reductions from coal-fired power plants, *Environ. Res. Lett.*, 10, DOI: 10.1088/1748-
449 9326/10/11/114015, 2015.

450 Wang, S., Xing, J., Jang, C., Zhu, Y., Fu, J., and Hao, J.: Impact Assessment of Ammonia Emissions on Inorganic Aerosols
451 in East China Using Response Surface Modeling Technique, *Environ. Sci. Technol.*, 45, 9293-9300, DOI:
452 10.1021/es2022347, 2011.

453 Wang, Z., Chien, C., and Tonnesen, G.: Development of a tagged species source apportionment algorithm to characterize
454 three-dimensional transport and transformation of precursors and secondary pollutants, *J. Geophys. Res.*, 114, DOI:
455 10.1029/2008JD010846, 2009.

456 Whitten, G., Heo, G., Kimura, Y., McDonald-Buller, E., Allen, D., Carter, W. P. L., and Yarwood, G.: A new condensed
457 toluene mechanism for Carbon Bond: CB05-TU, *Atmos. Environ.*, 44, 5346-5355, DOI:
458 10.1016/j.atmosenv.2009.12.029, 2010.

459 Wu, S., Hu, J., Zhang, Y., and Aneja, V. P.: Modeling atmospheric transport and fate of ammonia in North Carolina-Part
460 II: Effect of ammonia emissions on fine particulate matter formation, *Atmos. Environ.*, 42, 3437-3451, DOI:
461 10.1016/j.atmosenv.2007.04.022, 2008.

462 Xia, Y., Zhao, Y., and Nielsen, C. P.: Benefits of China's efforts in gaseous pollutant control indicated by the bottom-up
463 emissions and satellite observations 2000-2014, *Atmos. Environ.*, 136, 43-53, DOI: 10.1016/j.atmosenv.2016.04.013,
464 2016.

465 Xu, P., Liao, Y. J., Lin, Y. H., Zhao, C. X., Yan, C. H., Cao, M. N., Wang, G. S., and Luan, S. J.: High-resolution inventory
466 of ammonia emissions from agricultural fertilizer in China from 1978 to 2008, *Atmos. Chem. Phys.*, 16, 1207-1218,
467 DOI: 10.5194/acpd-15-25299-2015, 2016.

468 Yang, Y., Wilkinson, J., and Russell, A.: Fast, Direct Sensitivity Analysis of Multi-Dimensional Photochemical Models,
469 *Environ. Sci. Technol.*, 31, 2859-2868, DOI: 10.1021/es970117w, 1997.

470 Zhao, Z., Bai, Z., Winiwarter, W., Kiesewetter, G., Heyes, C., and Ma, L.: Mitigating ammonia emission from agriculture
471 reduces PM_{2.5} pollution in the Hai River Basin in China, *Sci. Total Environ.*, 609, 1152-1160, DOI:
472 10.1016/j.scitotenv.2017.07.240, 2017.

473 Zhao, X. J., Zhao, P. S., Xu, J., Meng, W., Pu, W. W., Dong, F., He, D., and Shi, Q. F.: Analysis of a winter regional haze
474 event and Its formation mechanism in the North China Plain, *Atmos. Chem. Phys.*, 13, 5685-5696, DOI: 10.5194/acp-
475 13-5685-2013, 2013.

476 Zhang, J. K., Sun, Y., Liu, Z. R., Ji, D. S., Hu, B., Liu, Q., and Wang, Y. S.: Characterization of submicron aerosols during
477 a month of serious pollution in Beijing, 2013, *Atmos. Chem. Phys.*, 14, 2887-2903, DOI: 10.5194/acp-14-2887-2014,
478 2014.

479 Zhang, K., Ma, Y., Xin, J., Liu, Z., Ma, Y., Gao, D., Wu, J., Zhang, W., Wang, Y., and Shen, P.: The aerosol optical
480 properties and PM_{2.5} components over the world's largest industrial zone in Tangshan, North China, *Atmos. Res.*, 201,
481 226-234, DOI: 10.1016/j.atmosres.2017.10.025, 2018.

482 Zhang, L., Chen, Y., Zhao, Y., Henze, D., Zhu, L., Song, Y., Paulot, F., Liu, X., Pan, Y., Lin, Y., and Huang, B.: Agricultural
483 ammonia emissions in China: reconciling bottom-up and top-down estimates, *Atmos. Chem. Phys.*, 18, 339-355, DOI:
484 10.5194/acp-18-339-2018, 2018.

485 Pan, Y., Tian, S., Zhao, Y., Zhang, L., Zhu, X., Gao, J., Huang, W., Zhou, Y., Song, Y., Zhang, Q., and Wang, Y.: Identifying
486 ammonia hotspots in China using a national observation network, *Environ. Sci. Technol.*, doi:
487 10.1021/acs.est.7b05235, DOI: 10.1021/acs.est.7b05235, 2018.

488 Zhang, X., Wang, Y., Niu, Y., Zhang, X., Gong, S., Zhang, Y., and Sun, J.: Atmospheric aerosol compositions in China:
489 spatial/temporal variability, chemical signature, regional haze distribution and comparisons with global aerosols,
490 *Atmos. Chem. Phys.*, 12, 779-799, DOI: 10.5194/acp-12-779-2012, 2012.

491 Zhou, F., Ciais, P., Hayashi, K., Galloway, J., Kim, D., Yang, L., Li, S., Liu, B., Shang, Z., and Gao, S.: Re-estimating NH₃

emissions from Chinese cropland by a new nonlinear model, *Environ. Sci. Technol.*, 50, 564-572, DOI: 10.1021/acs.est.5b03156, 2016.

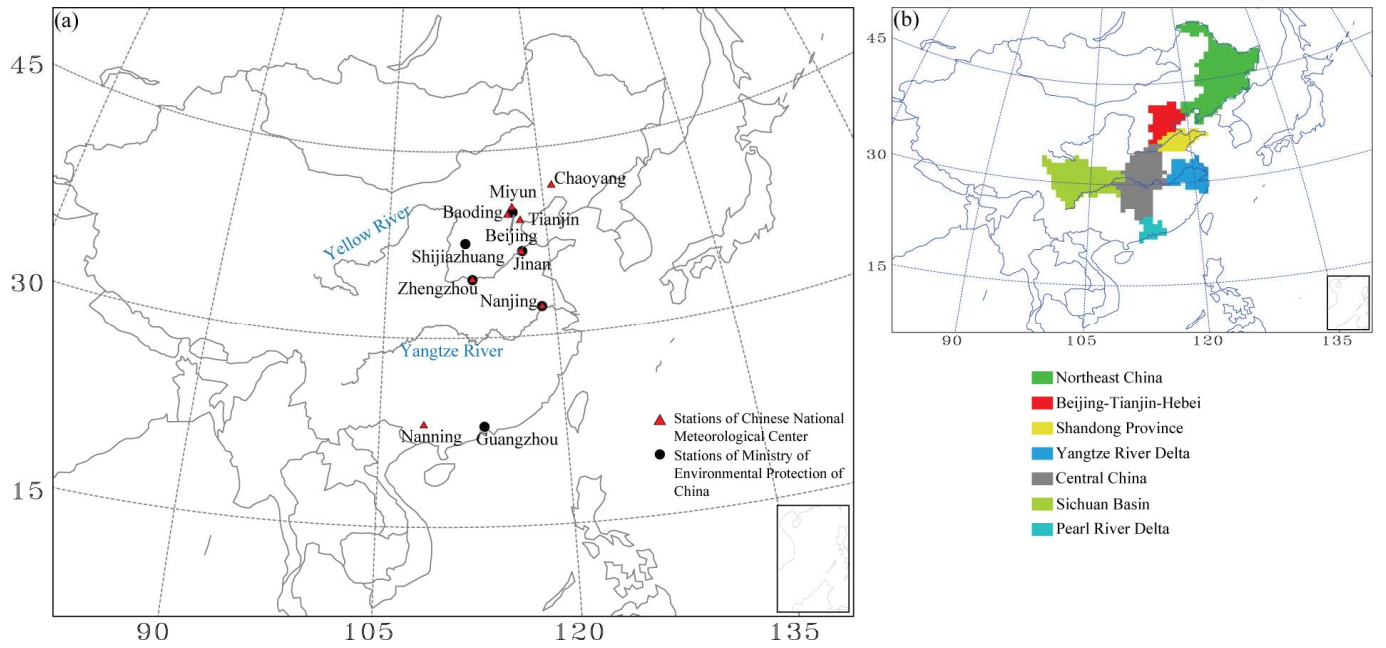
Zhou, Y., Cheng, S., Lang, J., Chen, D., Zhao, B., Liu, C., Xu, R., and Li, T., A comprehensive ammonia emission inventory with high-resolution and its evaluation in the Beijing–Tianjin–Hebei (BTH) region, China, *Atmos. Environ.*, 106, 305-317, DOI: 10.1016/j.atmosenv.2015.01.069, 2015.

Xu, P., Koloutsou-Vakakis, S., Rood, M., and Luan, S.: Projections of NH₃ emissions from manure generated by livestock production in China to 2030 under six mitigation scenarios, *Sci. Total Environ.*, 31, 78-86, DOI: 10.1016/j.scitotenv.2017.06.258, 2017.

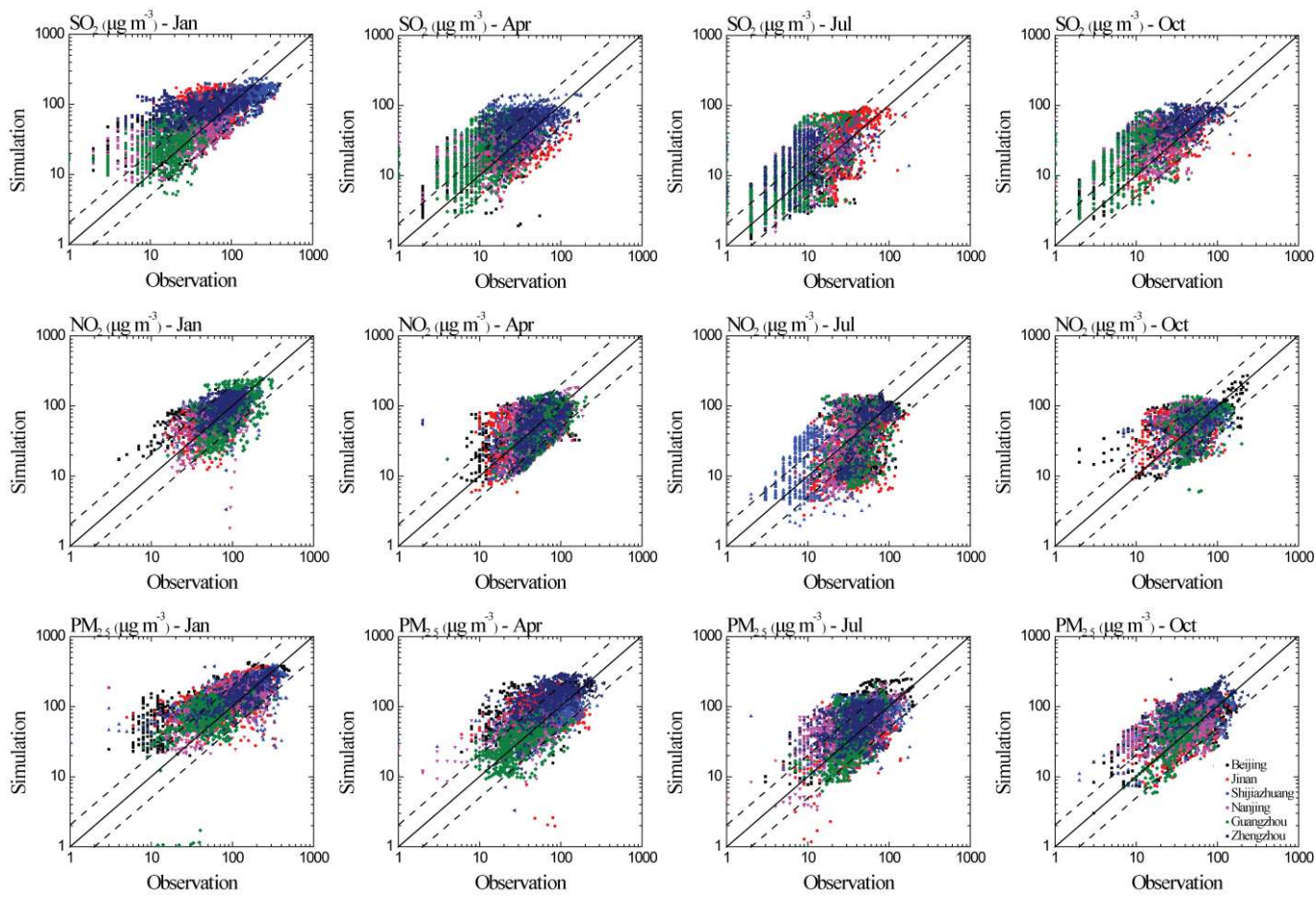
Zheng, B., Tong, D., Li, M., Liu, F., Hong, C., Geng, G., Li, H., Li, X., Peng, L., Qi, J., Yan, L., Zhang, Y., Zhao, H., Zheng, Y., He, K., and Zhang, Q.: Trends in China's anthropogenic emissions since 2010 as the consequence of clean air actions, *Atmos. Chem. Phys.*, 18, 14095-14111, DOI: 10.5194/acp-18-14095-2018, 2018.

Kurokawa, J., Ohara, T., Morikawa, T., Hanayama, S., Janssens-Maenhout, G., Fukui, T., Kawashima, K., and Akimoto, H.: Emissions of air pollutants and greenhouse gases over Asian regions during 2000-2008: Regional Emission inventory in ASia (REAS) version 2, *Atmos. Chem. Phys.*, 13, 11019-11058, DOI: 10.5194/acp-13-11019-2013, 2013.

Fu, X., Wang, S., Ran, L., Pleim, J., Cooter, E., Bash, J., Benson, V., and Hao, J.: Estimating NH₃ emissions from agricultural fertilizer application in China using the bi-directional CMAQ model coupled to an agro-ecosystem model, *Atmos. Chem. Phys.*, 15, 6637-6649, DOI: 10.5194/acp-15-6637-2015, 2015.



525
 526 Figure 1. Model domain used in this study and the geographic locations of Beijing-Tianjin-Hebei (BTH), Northeast
 527 China (NEC), Yangtze River Delta (YRD), Pearl River Delta (PRD), Sichuan Basin (SCB), Central China (CNC) and
 528 Shandong Province (SDP). The location of observation data was also shown in the model domain.
 529
 530
 531
 532
 533
 534
 535
 536
 537
 538
 539
 540
 541
 542
 543
 544
 545
 546
 547
 548
 549
 550
 551
 552
 553



554

555 Figure 2. The scatter plots between the modeled and the observed hourly SO₂, NO₂, and PM_{2.5} in January, April, July and
 556 October 2015. The solid lines are 1:1 and the dashed lines are 2:1 or 1:2.

557

558

559

560

561

562

563

564

565

566

567

568

569

570

571

572

573

574

575

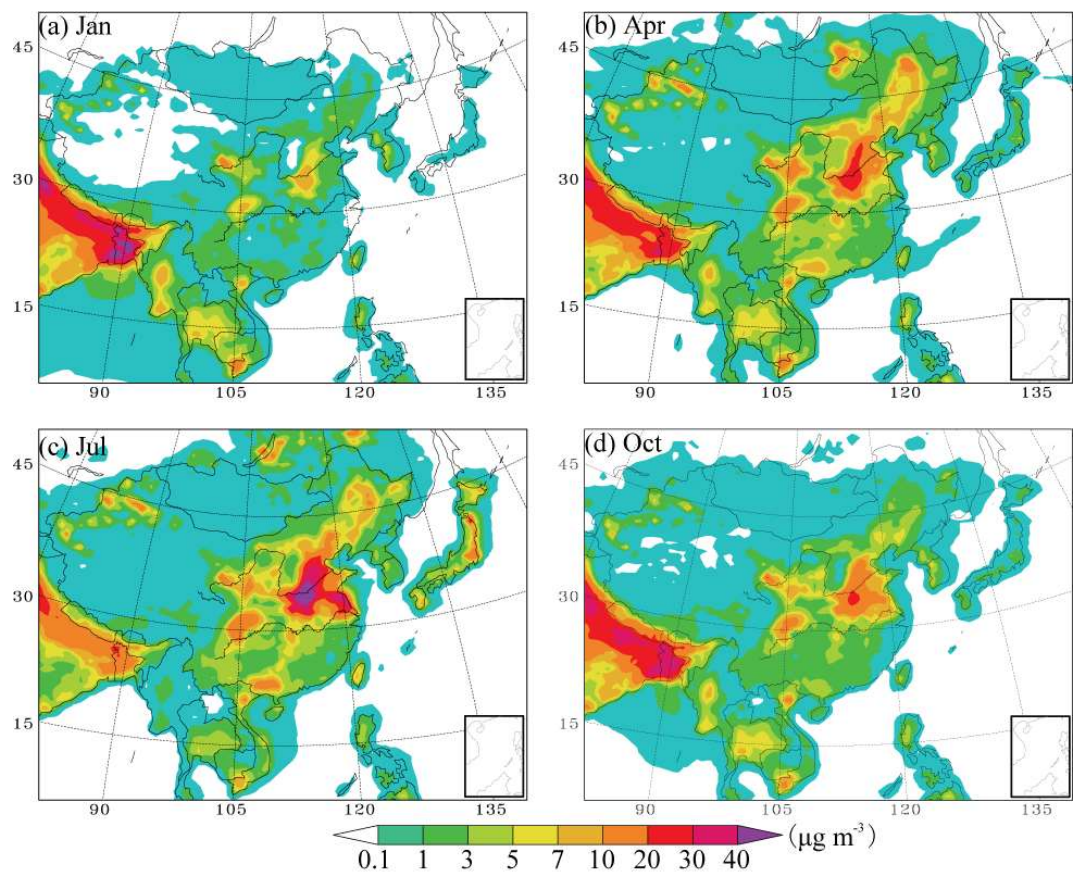


Figure 3. The horizontal distributions of the modeled monthly NH_3 mass concentration in January, April, July, and October in 2015.

576
 577
 578
 579
 580
 581
 582
 583
 584
 585
 586
 587
 588
 589
 590
 591
 592
 593
 594
 595
 596
 597
 598

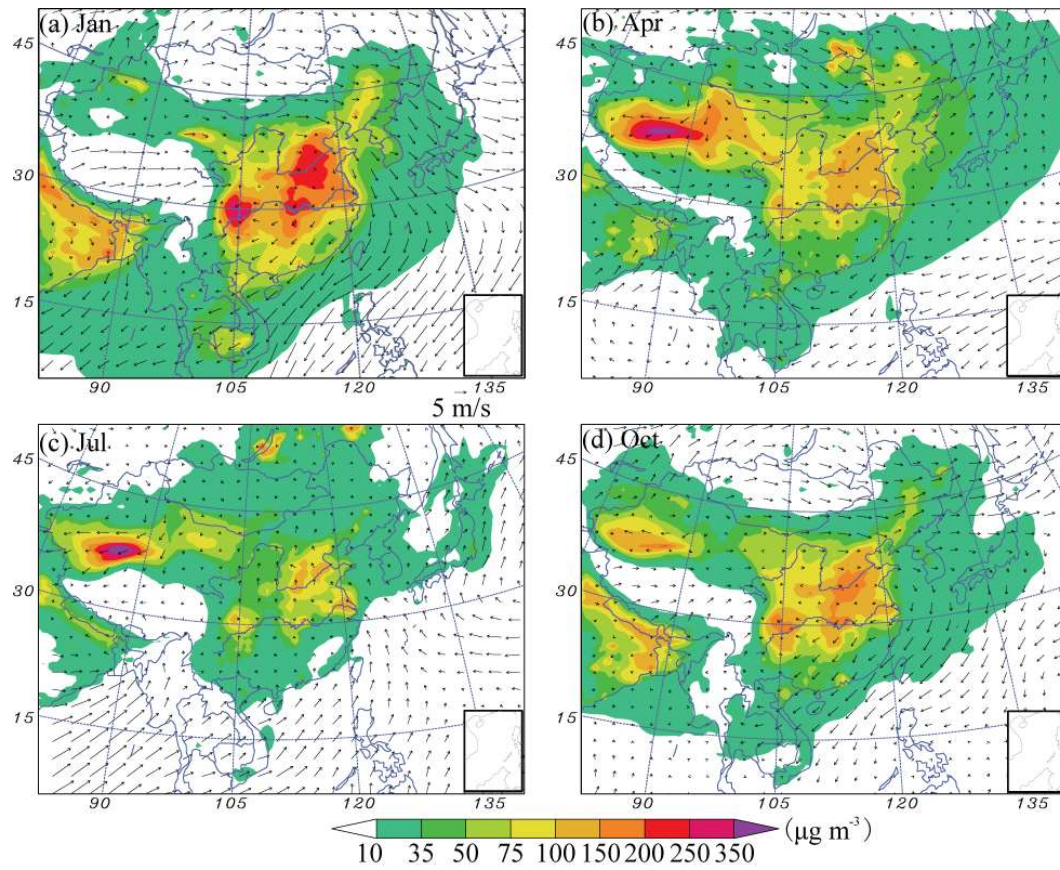


Figure 4. The horizontal distributions of the modeled monthly PM_{2.5} mass concentration in January, April, July, and October in 2015. Also shown are the surface wind field.

599
 600
 601
 602
 603
 604
 605
 606
 607
 608
 609
 610
 611
 612
 613
 614
 615
 616
 617
 618
 619
 620
 621
 622

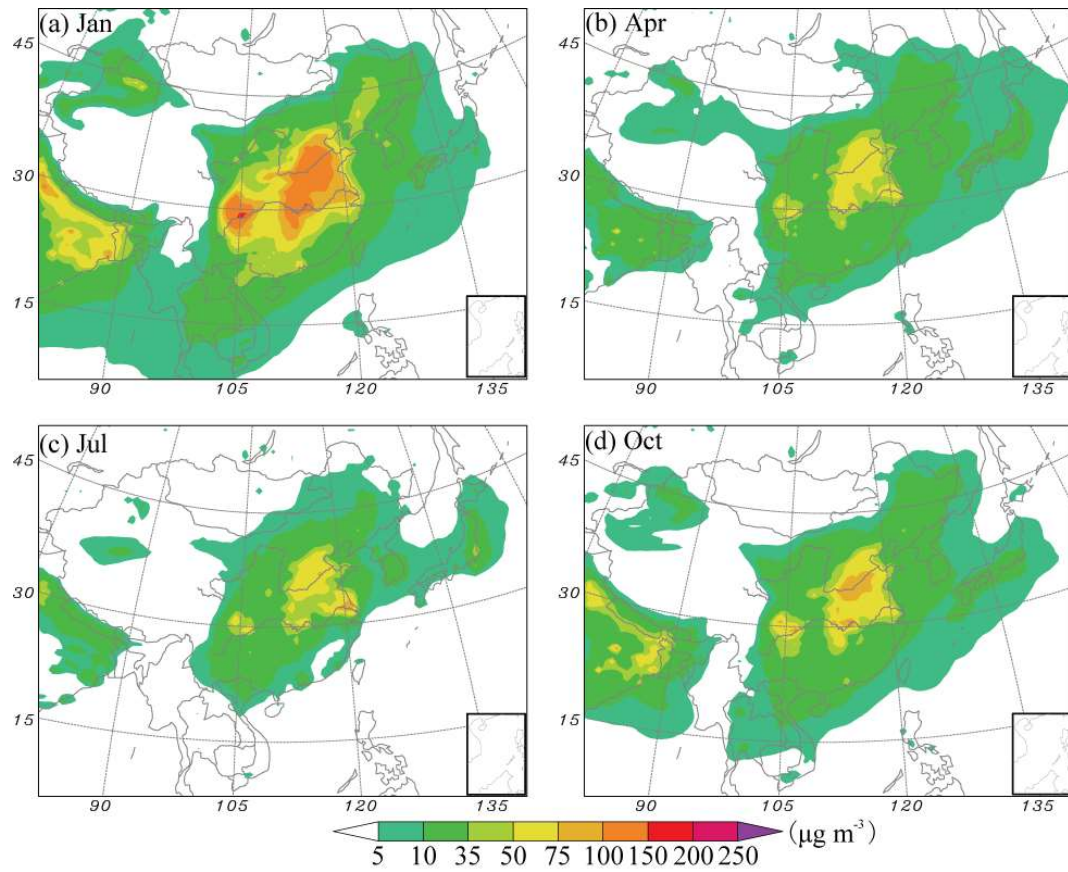
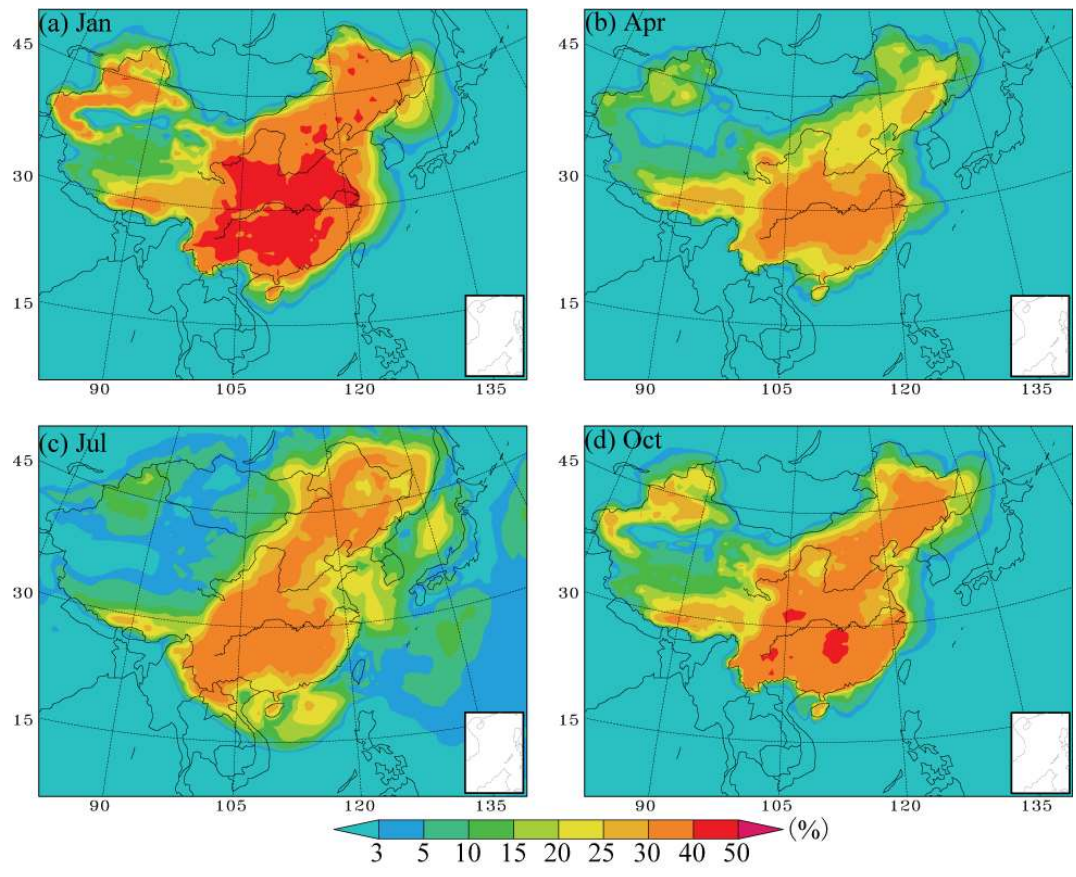


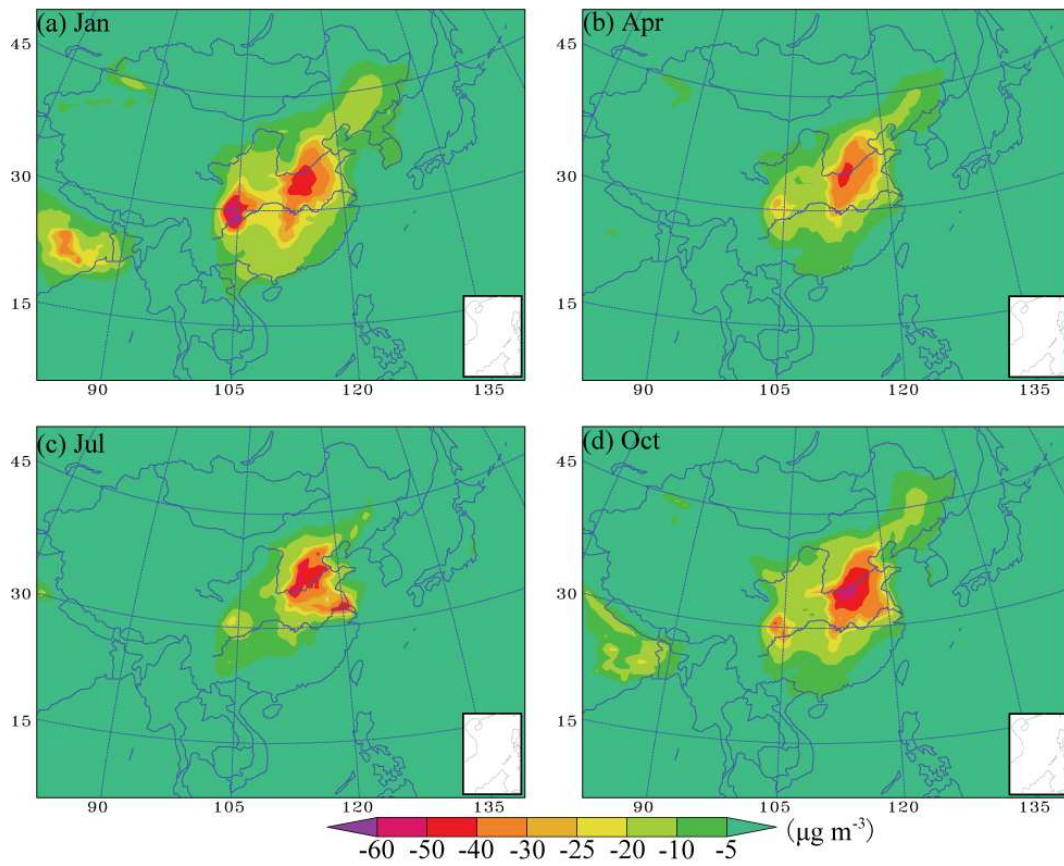
Figure 5. The horizontal distributions of the modeled monthly SNA mass concentration in January, April, July, and October in 2015.

623
 624
 625
 626
 627
 628
 629
 630
 631
 632
 633
 634
 635
 636
 637
 638
 639
 640
 641
 642
 643
 644
 645
 646



647
 648 Figure 6. The horizontal distributions of the contribution percentage of NH_3 emissions to SNA mass concentration (%) in
 649 January, April, July and October.
 650

651
 652
 653
 654
 655
 656
 657
 658
 659
 660
 661
 662
 663
 664
 665
 666
 667
 668
 669
 670



671
 672 Figure 7. The horizontal distributions of SNA mass concentration ($\mu\text{g m}^{-3}$) variation associated with agriculture NH_3
 673 removal in January, April, July and October.
 674
 675
 676
 677
 678
 679
 680
 681
 682
 683
 684
 685
 686
 687
 688
 689
 690
 691
 692
 693
 694

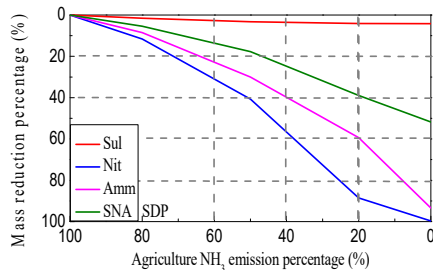
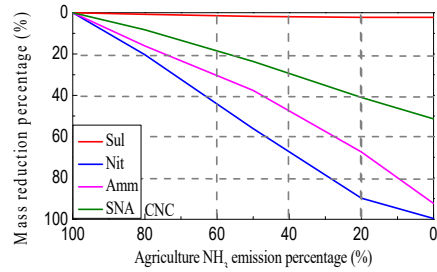
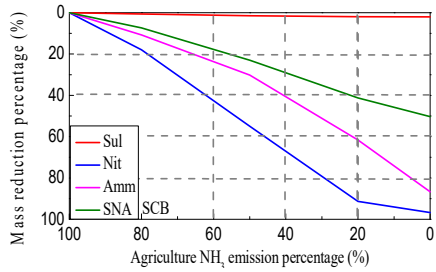
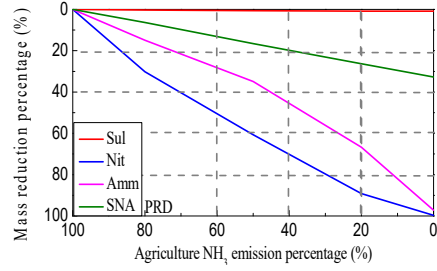
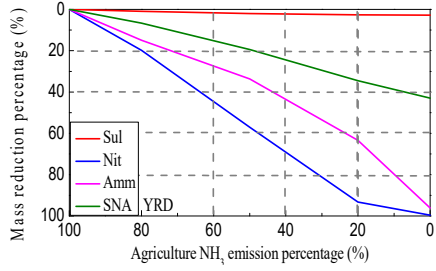
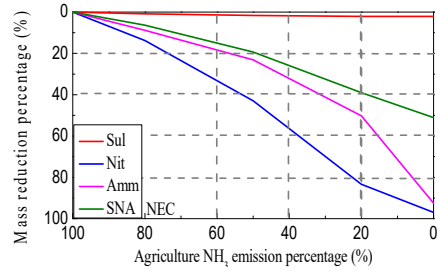
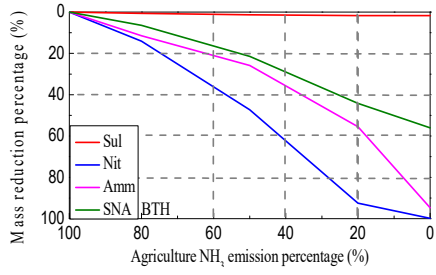


Figure 8. The variation (%) of sulfate, nitrate, ammonium, and SNA mass burden associated with the NH₃ emission reduction (%).

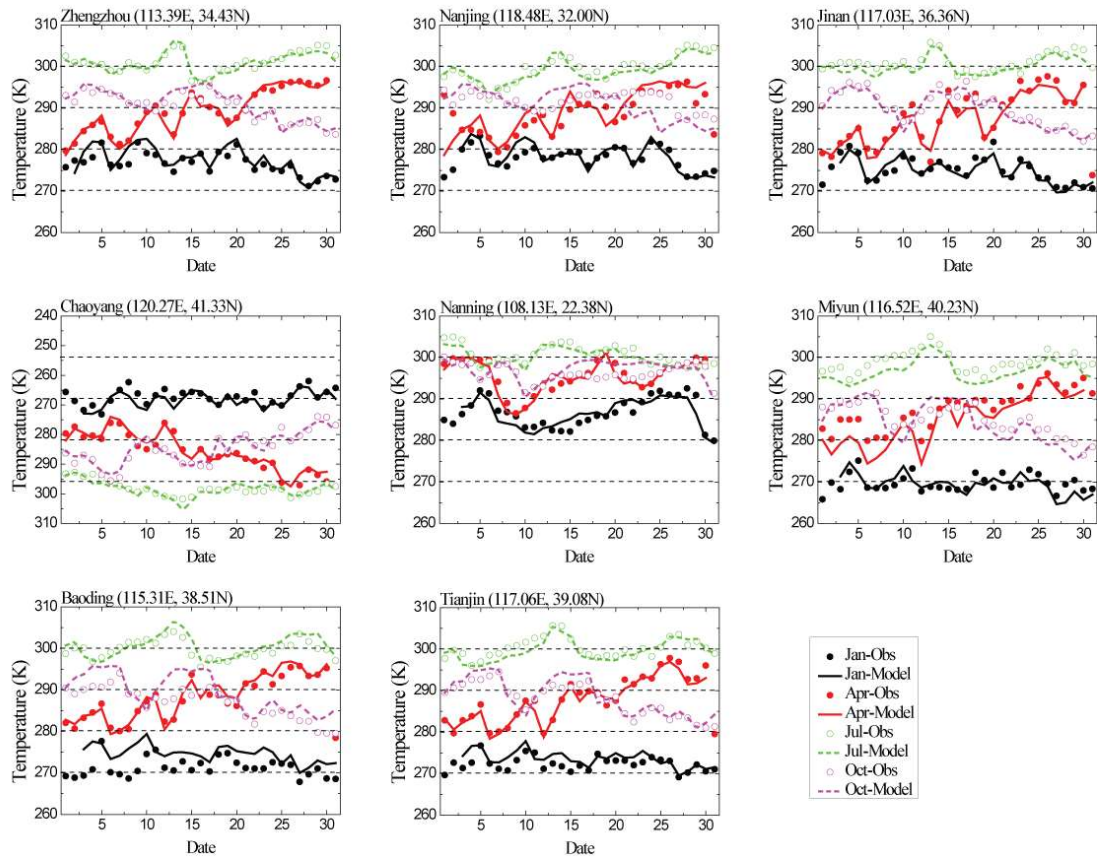


Figure A1. Observed and modeled daily average temperatures (K) in January, April, July and October 2015.

713
 714
 715
 716
 717
 718
 719
 720
 721
 722
 723
 724
 725
 726
 727
 728
 729
 730
 731
 732
 733
 734
 735

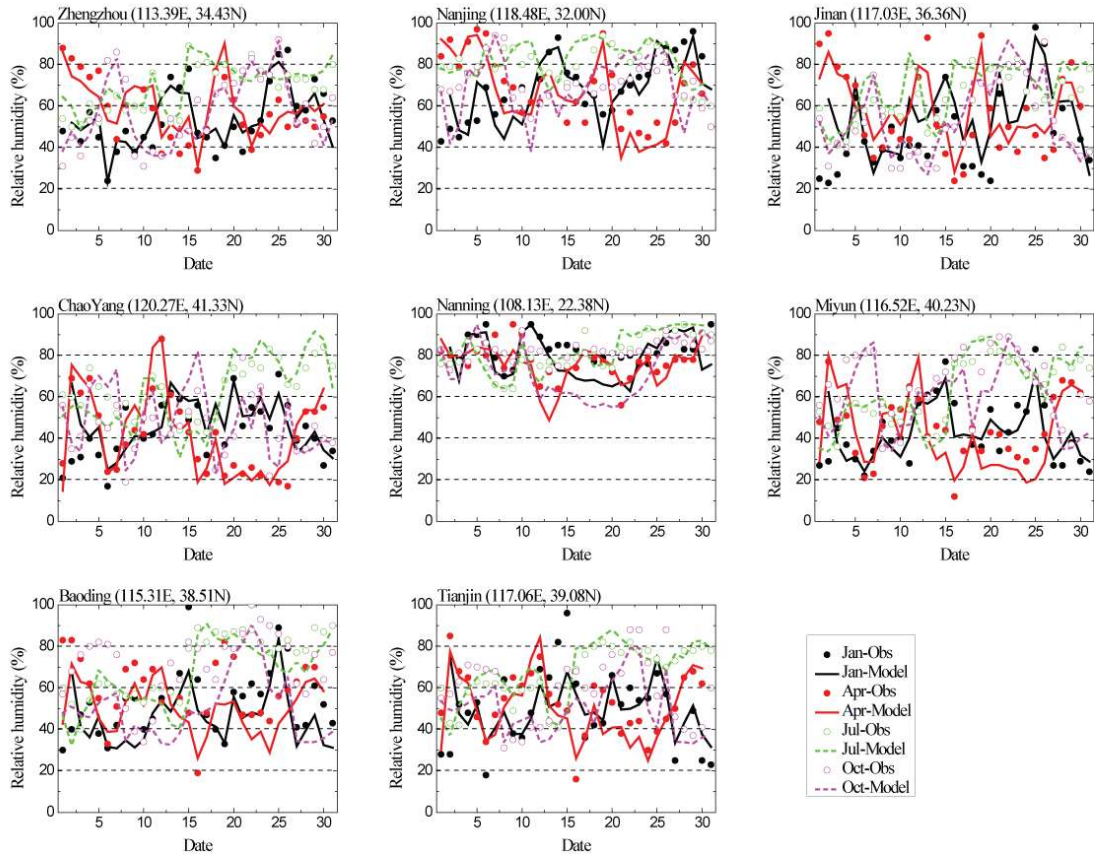


Figure A2. Same as Figure A1 but for relative humidity (%)

736
 737
 738
 739
 740
 741
 742
 743
 744
 745
 746
 747
 748
 749
 750
 751
 752
 753
 754
 755
 756
 757
 758
 759

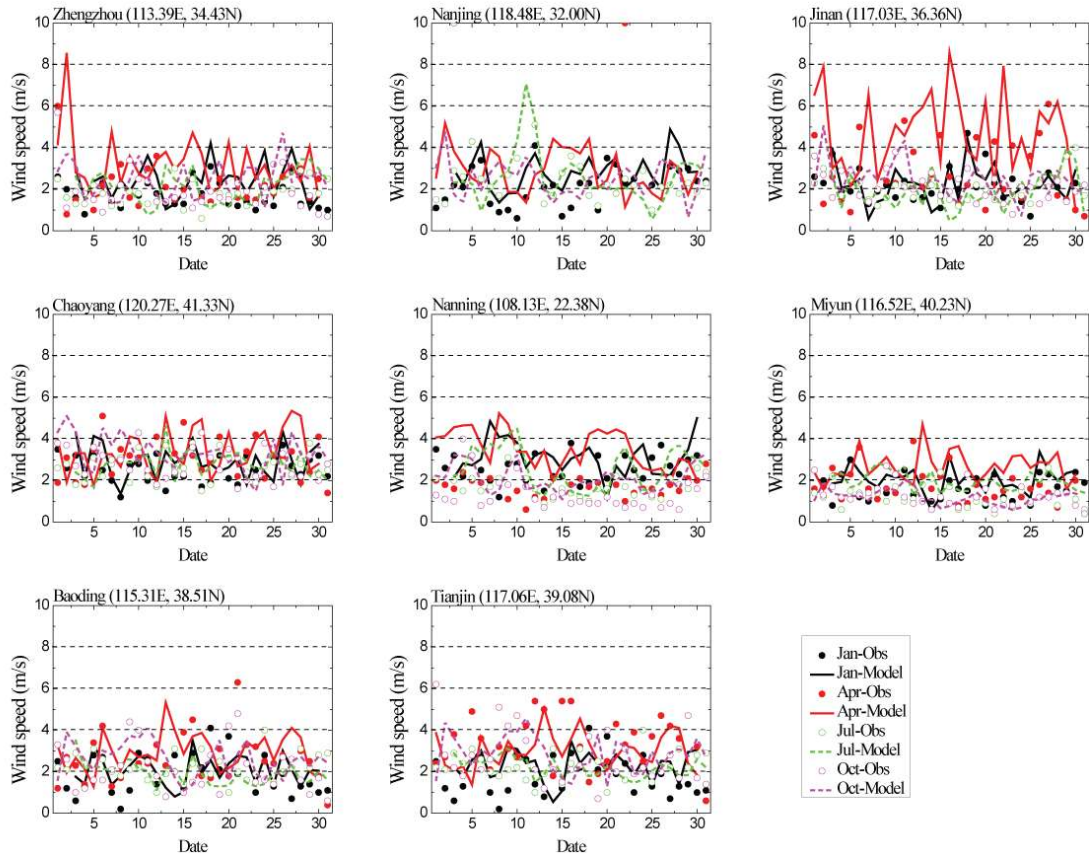


Figure A3. Same as Figure A1 but for wind speed (m s^{-1})

760
 761
 762
 763
 764
 765
 766
 767
 768
 769
 770
 771
 772
 773
 774
 775
 776
 777
 778
 779
 780
 781
 782
 783

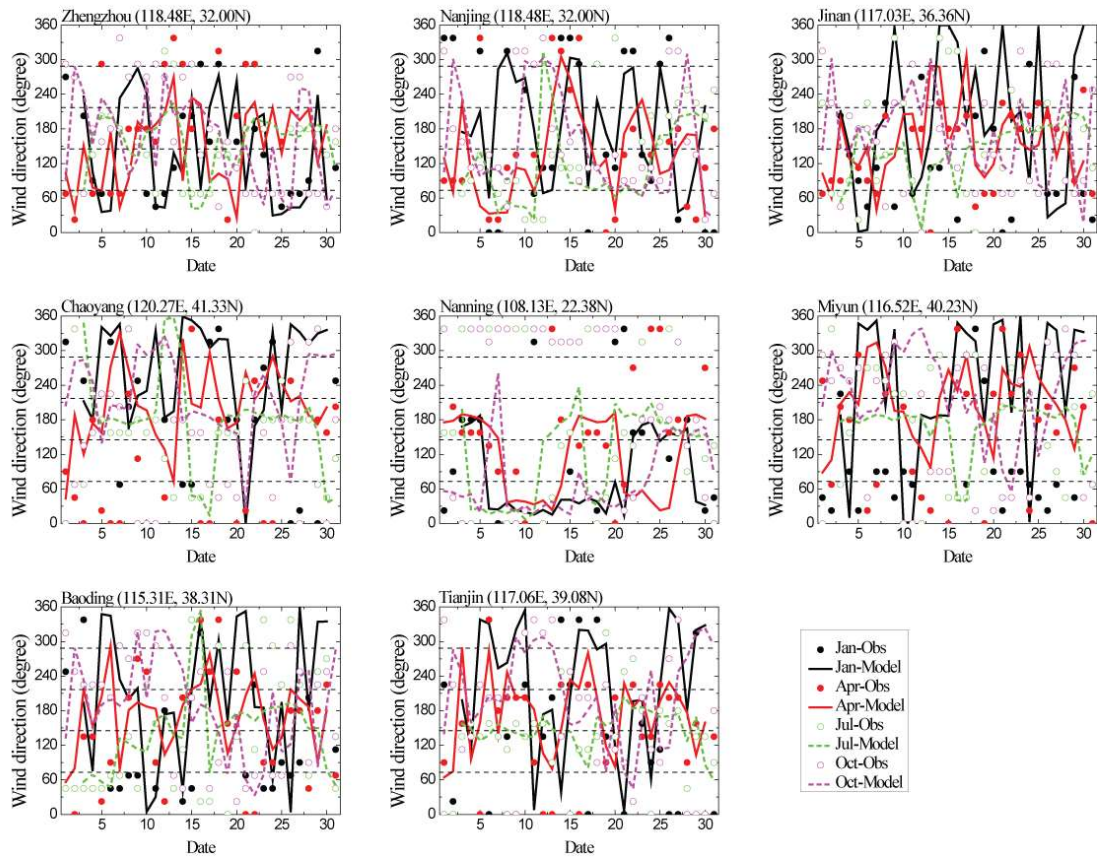
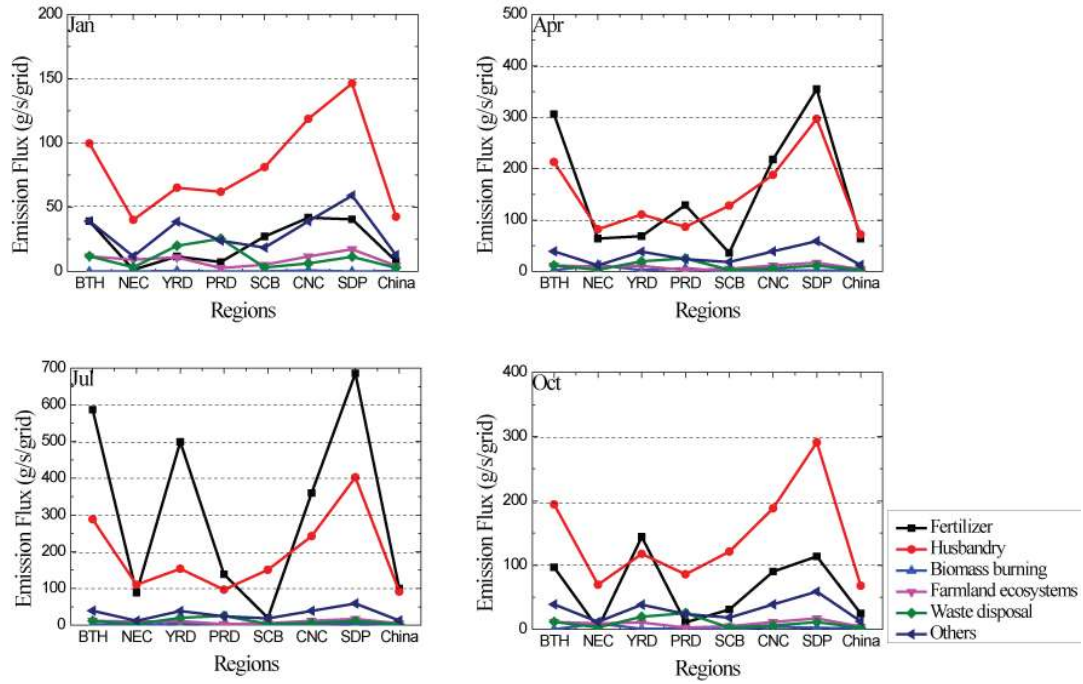


Figure A4. Same as Figure A1 but for daily maximum wind direction (degree)

784
785
786
787
788
789
790
791
792
793
794
795
796
797
798
799
800
801
802
803
804
805
806
807



808
 809 Figure A5. The regional average NH₃ emission flux (g/s/grid) of different agriculture sectors over each region in January,
 810 April, July and October.
 811
 812
 813
 814
 815
 816
 817
 818
 819
 820
 821
 822
 823
 824
 825
 826
 827
 828
 829
 830
 831
 832
 833
 834
 835

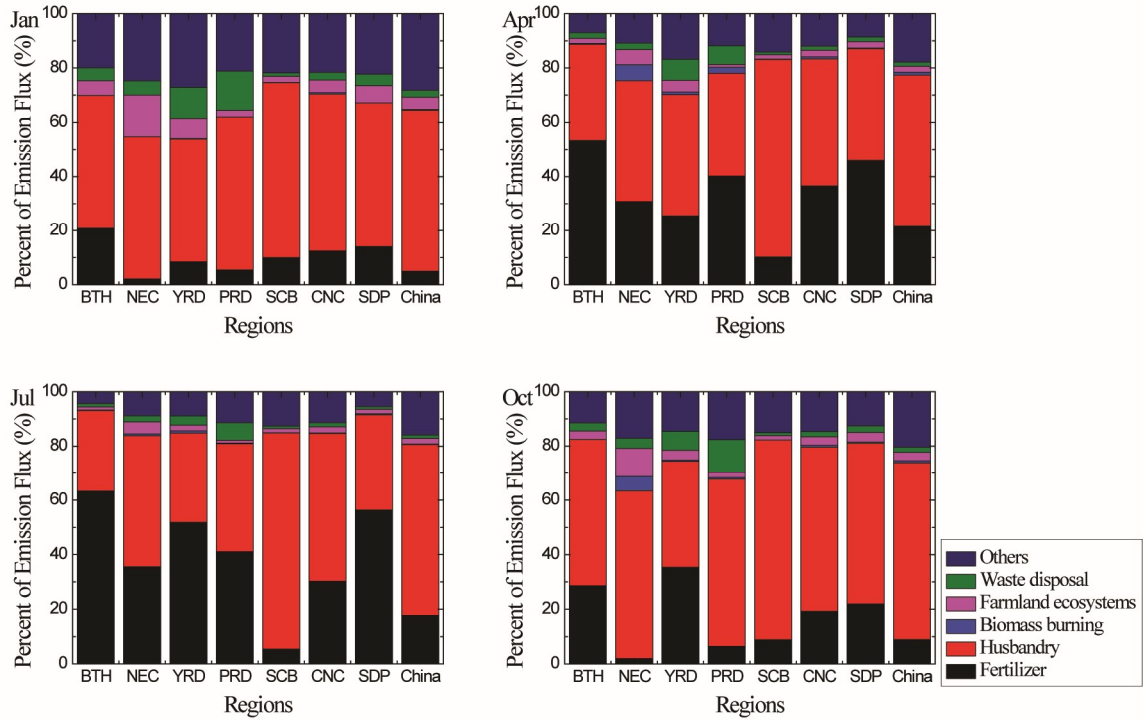


Figure A6. The percent (%) of different NH₃ emission sectors over each region in January, April, July and October.

836
837
838
839
840
841
842
843
844
845
846
847
848
849
850
851
852
853
854
855
856
857
858
859
860
861
862

863
864
865
866
867
868
869
870
871
872
873
874
875
876
877
878
879
880
881
882
883
884
885
886
887
888
889
890
891
892
893
894
895
896
897
898
899
900
901

Table 1. Statistical summary of the comparisons of the monthly average PM_{2.5} between simulation and observation

	N^a	O^b	M^c	σ_o^d	σ_m^e	R^f
Jan	4464	106.5	126.9	84.5	76.2	0.74
Apr	4320	64.6	76.8	44.7	56.8	0.66
Jul	4464	49.2	42.1	32.3	41.4	0.58
Oct	4464	58.2	68.1	35.5	46.9	0.61

- ^a Number of samples
- ^b Total mean of observation
- ^c Total mean of simulation
- ^d Standard deviation of observation
- ^e Standard deviation of simulation
- ^f Correlation coefficient between daily observation and simulation

902
903
904
905
906
907
908
909
910
911
912
913
914
915
916
917
918
919
920
921
922
923
924
925
926
927
928
929
930
931
932
933
934
935
936
937
938
939
940

Table 2. Statistical summary of the comparisons of the monthly average NO₂ between simulation and observation

	<i>N</i>	<i>O</i>	<i>M</i>	σ_o	σ_m	<i>R</i>
Jan	4464	79.7	87.1	39.8	39.0	0.60
Apr	4320	53.7	55.5	29.9	32.5	0.59
Jul	4464	43.6	40.1	25.8	30.0	0.51
Oct	4464	53.6	61.3	32.0	31.7	0.54

941
942
943
944
945
946
947
948
949
950
951
952
953
954
955
956
957
958
959
960
961
962
963
964
965
966
967
968
969
970
971
972
973
974
975
976
977
978
979

Table 3. Statistical summary of the comparisons of the monthly average SO₂ between simulation and observation

	<i>N</i>	<i>O</i>	<i>M</i>	σ_o	σ_m	<i>R</i>
Jan	4464	61.0	71.5	61.3	47.5	0.63
Apr	4320	24.8	35.7	26.5	24.6	0.52
Jul	4464	13.9	22.4	15.1	19.1	0.46
Oct	4464	21.2	36.8	20.2	21.8	0.50

980

981

Table 4. The regional percent (%) of Tagr NH₃ contribution to sulfate, nitrate, ammonium, and SNA mass concentration.

		Sulfate	Nitrate	Ammonium	SNA	PM _{2.5}
BTH	Jan	0.9	4.5	98.0	39.7	19.3
	Jul	1.0	9.3	75.9	28.1	20.6
	Annual	1.1	8.0	83.3	23.1	15.5
NEC	Jan	0.6	3.2	94.0	34.4	18.6
	Jul	0.8	6.7	83.5	27.9	16.1
	Annual	1.0	5.6	83.7	22.5	14.3
YRD	Jan	0.9	5.8	99.2	40.9	22.5
	Jul	0.5	8.1	68.7	24.0	15.4
	Annual	1.0	7.4	85.7	23.6	15.3
PRD	Jan	0.8	5.0	98.1	40.2	20.4
	Jul	1.4	4.7	85.3	27.7	15.9
	Annual	0.9	5.8	90.6	24.5	14.2
SCB	Jan	0.6	3.7	97.0	37.4	17.9
	Jul	0.7	5.6	95.9	31.5	19.5
	Annual	0.7	5.1	93.9	21.6	13.0
CNC	Jan	0.9	4.9	99.2	42.6	20.6
	Jul	0.9	6.7	88.9	33.7	22.0
	Annual	0.9	6.0	92.8	26.1	17.5
SDP	Jan	0.7	4.9	98.3	39.2	21.0
	Jul	0.7	8.3	67.0	23.5	16.6
	Annual	0.9	7.1	80.5	21.6	15.1
China	Jan	2.4	9.3	92.3	34.4	21.4
	Jul	2.2	10.4	90.9	25.1	16.4
	Annual	2.2	10.1	87.6	29.0	16.0

982

983

984

985

986

987

988

989

990

991

992

993

994

995

996

997

998

999
1000
1001

Table 5. The variation percent (%) of sulfate, nitrate, ammonium, and SNA mass concentration associated with agriculture NH₃ removal.

		Sulfate	Nitrate	Ammonium	SNA	PM _{2.5}
BTH	Jan	0.5	99.8	96.2	51.9	37.8
	Jul	1.0	99.6	95.0	47.0	39.2
	Annual	0.7	99.8	94.7	49.4	38.5
NEC	Jan	0.7	99.2	96.4	60.9	39.2
	Jul	0.8	94.5	91.5	37.0	27.8
	Annual	0.7	96.9	92.5	48.9	34.5
YRD	Jan	2.7	99.4	96.0	52.6	32.2
	Jul	7.2	99.0	96.8	44.9	37.6
	Annual	5.0	99.2	96.1	48.8	36.9
PRD	Jan	3.6	99.8	97.2	50.3	31.5
	Jul	0.4	92.7	97.4	30.3	24.1
	Annual	2.0	96.2	97.2	40.3	27.8
SCB	Jan	4.9	94.1	80.3	57.6	41.7
	Jul	0.2	99.3	92.5	42.0	28.3
	Annual	2.6	96.7	85.9	49.8	35.0
CNC	Jan	3.1	99.1	92.2	56.7	41.5
	Jul	0.7	99.3	96.0	45.1	37.2
	Annual	1.9	99.2	92.3	50.9	39.4
SDP	Jan	1.7	99.8	95.8	47.7	38.0
	Jul	3.6	99.2	93.6	45.6	37.3
	Annual	2.7	99.5	93.4	46.6	37.6
China	Jan	2.6	93.9	86.3	54.8	39.5
	Jul	0.6	97.7	87.8	36.7	27.5
	Annual	1.6	95.8	86.9	45.7	32.9

1002
1003
1004
1005
1006