



1	Long-term trends and drivers of aerosol pH in eastern China
2	Min Zhou ^{1,2} , Guangjie Zheng ³ , Hongli Wang ¹ , Liping Qiao ¹ , Shuhui Zhu ¹ , Dandan Huang ¹ , Jingyu An ¹ ,
3	Shengrong Lou ¹ , Shikang Tao ¹ , Qian Wang ¹ , Rusha Yan ¹ , Yingge Ma ¹ , Changhong Chen ¹ , Yafang Cheng ³ ,
4	Hang Su*,1,4, Cheng Huang¹
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7	¹ State Environmental Protection Key Laboratory of the Cause and Prevention of Urban Air Pollution
8	Complex, Shanghai Academy of Environmental Sciences, Shanghai200233, China
9	² School of Atmospheric Sciences, Nanjing University, Nanjing210023, China
10	³ Minerva Research Group, Max Planck Institute for Chemistry, Mainz 55128, Germany
11	⁴ Multiphase Chemistry Department, Max Planck Institute for Chemistry, Mainz 55128, Germany
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13	*Corresponding author: Hang Su (<u>h.su@mpic.de</u>)
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Abstract

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21 Aerosol acidity plays a key role in regulating the chemistry and toxicity of atmospheric aerosol particles. 22 The trend of aerosol pH and its drivers are crucial in understanding the multiphase formation pathways 23 of aerosols. Here, we reported the first trend analysis of aerosol pH from 2011 to 2019 in eastern China. 24 The implementation of the Air Pollution Prevention and Control Action Plan leads to -35.8%, -25 37.6%, -9.6%, -81.0% and 1.2% changes of PM_{2.5}, SO₄²⁻, NH_x, NVCs and NO₃⁻ in YRD during this 26 period. Different from the fast changes of aerosol compositions due to the implementation of the Air 27 Pollution Prevention and Control Action Plan, aerosol pH shows a moderate change of -0.24 unit over the 9 years. Besides the multiphase buffer effect, the opposite effects of SO₄²⁻ and non-volatile cations 28 29 changes play key roles in determining the moderate pH trend, contributing to a change of +0.38 and 30 -0.35 unit, respectively. Seasonal variations in aerosol pH were mainly driven by the temperature, while 31 the diurnal variations were driven by both temperature and relative humidity. In the future, SO₂, NO_x 32 and NH₃ emissions are expected to be further reduced by 86.9%, 74.9% and 41.7% in 2050 33 according to the best health effect pollution control scenario (SSP1-26-BHE). The corresponding 34 aerosol pH in eastern China is estimated to increase by ~0.9, resulting in 8% more NO₃- and 35% less 35 NH₄⁺ partitioning/formation in the aerosol phase, which suggests a largely reduced benefit of NH₃ and 36 NO_x emission control in mitigating haze pollution in eastern China.

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1 Introduction

- Aerosol acidity is an important parameter in atmospheric chemistry studies. It affects the particle mass and chemical composition by regulating the reactions of aerosols, and is closely associated with human health, ecosystems and climate(Li et al., 2017; Nenes et al., 2020b; Pye et al., 2020; Su et al., 2020).

 Aerosol acidity has attracted increasing concern in recent years because of its impacts on the thermodynamics of gas-particle partitioning, pH-dependent condensed-phase reactions and trace metal solubility(Cheng et al., 2016; Fang et al., 2017; Guo et al., 2017b; Guo et al., 2016; He et al., 2018; Song et al., 2018; Weber et al., 2016; Su et al., 2020).
- 46 Aerosol pH is normally estimated using thermodynamic models, such as E-AIM(Clegg et al., 1998)





47 and ISORROPIA II, due to the limitations of direct aerosol pH measurement techniques(Fountoukis and 48 Nenes, 2007; Hennigan et al., 2015). The global distribution of aerosol pH generally ranges from 1 to 49 6(Pye et al., 2020; Zheng et al., 2020; Su et al., 2020). Aerosols in the United States are highly acidic, 50 with pH values of approximately 1-2(Guo et al., 2015; Nah et al., 2018; Pye et al., 2018; Zheng et al., 51 2020). Aerosols in mainland China and Europe have similar average aerosol acidity levels (pH = 2.5-6)(Guo et al., 2018; Jia et al., 2018; Masiol et al., 2020; Shi et al., 2019; Tan et al., 2018; Wang et al., 52 53 2019; Zheng et al., 2020). 54 Aerosol pH exhibits notable spatial and temporal variability, which is affected by changes in factors 55 such as temperature, relative humidity (RH), and aerosol compositions(Pye et al., 2018; Nenes et al., 56 2020a; Tao et al., 2020; Zheng et al., 2020). Very few studies have investigated the trend and spatial variability of aerosol pH and its drivers. Weber et al. (Weber et al., 2016) showed that aerosols tend to 57 58 remain highly acidic upon the reduction of SO₄²⁻ during summertime in the southeastern United States. 59 Based on the 10-year observations in six Canadian sites, Tao and Murphy (Tao and Murphy, 2019) 60 suggested that meteorological parameters are more important than the chemical compositions in controlling aerosol pH variations. Zheng et al., (Zheng et al., 2020) found that aerosol liquid water content 61 62 (ALWC) and temperature are the main factors that contribute to the pH difference between the wintertime 63 North China Plain and summertime southeastern United States, whereas the change of chemical composition only plays a minor role (15%). In China, the trend of aerosol pH and its drivers remain 64 poorly understood, especially in recent years when the emissions and aerosol compositions undergo 65 66 substantial changes. 67 To tackle severe particulate matter pollution in China, the Chinese government released the Air 68 Pollution Prevention and Control Action Plan (hereinafter referred to as the Action Plan) in September 69 2013, which is the first plan to specify air quality goals in China(Cai et al., 2017; Liu et al., 2018; Zheng 70 et al., 2018). The implementation of the Action Plan has led to significant changes in the concentrations 71 and chemical characteristics of fine particulate (PM2.5). Aerosol pH may change due to the significant 72 changes of the chemical composition in PM2.5, which may feedback to the multiphase formation 73 pathways of aerosols such as sulfate, nitrate and ammonium(Cheng et al., 2016). 74 In this study, we performed a comprehensive analysis of the long-term trends of aerosol pH and its 75 drivers in the Yangtze River Delta of eastern China. In this study, a thermodynamic model, ISORROPIA





II (version 2.1)(Fountoukis and Nenes, 2007) was applied to estimate the pH based on 9-year continuous online measurements of PM_{2.5} compositions at an urban site in Shanghai. The main purposes of this study were to: (1) characterizing the long-term trend in aerosol pH; (2) investigate the seasonal and diurnal variations of aerosol pH and the main factors that affect these changes and (3) predict further pH under different emission control scenarios and its impact on the formation of ammonium and nitrate. The results presented here may help to advance our understanding of aerosol chemistry in China and support the development of effective pollution control strategy.

2 Material and Methods

2.1 Ambient measurements

The observational site in this study is located in the Shanghai Academy of Environmental Sciences (SAES, 31°10′N, 121°25′E), a mixed commercial and residential district in the southwest central urban area of Shanghai (Fig. S1). In the absence of a significant nearby industrial source, this sampling site can be regarded as a representative urban area influenced by a wide mixture of emission sources. A detailed description can be found in previous studies(Qiao et al., 2014; Zhou et al., 2016).

The sampling was conducted from 2011 to 2019. Hourly mass concentrations of water-soluble gases (HCl, HNO₂, SO₂, HNO₃, NH₃) and major water-soluble inorganic ions in PM_{2.5}, including SO₄²⁻, nitrate (NO₃⁻), chloride (Cl⁻), ammonium (NH₄⁺), sodium (Na⁺), potassium (K⁺), calcium (Ca²⁺) and magnesium (Mg²⁺), were measured using an on-line analyser to monitor aerosols and gases (MARGA ADI 2080, Applikon Analytical B.V). The details of measurements were given in Qiao et al. (Qiao et al., 2014). To better track the retention time changes of different ion species and ensure their concentrations to be measured successfully, an internal calibration was conducted every hour with Lithium Bromide (LiBr) standard solution (Qiao et al., 2014; Zhou et al., 2016). In addition, cleaning the sampling system of MARGA and the multi-points calibrations with the standard solutions were performed every three months to ensure the accuracy of MARGA. Figure S2 compares the sum of SO_4^{2-} , NO_3^- and Cl^- with the sum of NH_4^+ , Na^+ , K^+ , Ca^{2+} and Mg^{2+} in neq/m³ to check the charge balance. The correlation between cation and anion was strong (R²=0.94), with a slope of 1.00, indicating that these ion species represented the major ions in $PM_{2.5}$, and the anion and cation were charge balanced. In previous studies, intercomparison experiments between MARGA and filter-based method have been carried out, and the





- data measured by MARGA show acceptable accuracy and precision(Rumsey et al., 2014; Huang et al.,
- 105 2014). The mass concentrations of PM_{2.5} were simultaneously measured using an on-line PM monitor
- 106 (FH 62 C14 series, Thermo Fisher Scientific) using beta attenuation techniques at a time resolution of 5
- min. The temperature and RH were also measured at a time resolution of 1 min.

2.2 Aerosol pH prediction

- 109 The aerosol pH was predicted using the ISORROPIA II thermodynamic model(Fountoukis and Nenes,
- 110 2007). ISORROPIA II can calculate the equilibrium H_{air}^+ and aerosol liquid water content of inorganic
- 111 material $(ALWC_i)$ by inputting the concentrations of the total SO_4^{2-} (TH₂SO₄, replaced by observed
- 112 SO₄²⁻), total NO₃⁻ (TNO₃, gas HNO₃ plus particle NO₃⁻), total ammonia (NH_x, gas NH₃ plus particle
- 113 NH₄⁺), total Cl⁻ (TCl, replaced by observed Cl⁻ due to the low concentration and measurement
- uncertainties of HCl)(Rumsey et al., 2014), non-volatile cations (NVCs, observed Na⁺, K⁺, Ca²⁺, Mg²⁺)
- and meteorological parameters, including temperature and RH(Guo et al., 2016). H_{air}^+ and $ALWC_i$ are
- then used to obtain the $PM_{2.5}$ pH by Eq. (1).

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$$pH = -log_{10}H_{aq}^{+} \cong -log_{10}\frac{1000H_{air}^{+}}{ALWC_{l} + ALWC_{o}} \cong -log_{10}\frac{1000H_{air}^{+}}{ALWC_{l}},$$
 (1)

- where H_{aq}^+ is the H⁺ concentration in solution (mol/L), H_{air}^+ is the H⁺ loading for an air sample ($\mu g/m^3$)
- and ALWCi and ALWCo are the aerosol liquid water contents of inorganic and organic species,
- respectively ($\mu g/m^3$). ALWC_o was calculated by Eq. (2) (Guo et al., 2015).

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$$ALWC_o = \frac{m_{org}\rho_w}{\rho_{org}} \frac{k_{org}}{(\frac{1}{PH} - 1)} , \qquad (2)$$

- where m_{org} is the mass concentration of organic aerosol, ρ_w is the density of water (ρ_w =1.0g/cm³),
- 123 ρ_{org} is the density of organics (ρ_{org} =1.4g/cm³)(Guo et al., 2015), and k_{org} is the hygroscopicity
- parameter of organic aerosol ($k_{org} = 0.087$)(Li et al., 2016). The annual $ALWC_o$ calculated were 1.4–
- 125 2.5μg/m³, only accounting for 4.3%–7.5% of the total aerosol liquid water content. The pH predictions
- in previous studies were insensitive to $ALWC_o$ unless the mass fraction of $ALWC_o$ to the total
- 127 aerosol liquid water content was close to unity(Guo et al., 2015). The use of ALWC_i to predict pH is
- therefore fairly accurate and common(Battaglia et al., 2017; Ding et al., 2019). In this study,
- 129 ISORROPIA II was run in the forward mode and 'metastable' state. Calculations using total (gas and
- 130 aerosol) measurements in the forward mode are less affected by measurement errors(Hennigan et al.,





131 2015; Song et al., 2018). A detailed description of the pH calculations can be found in previous 132 studies(Guo et al., 2017a; Guo et al., 2015; Song et al., 2018). 133 Figure S3 compares the predicted and measured concentrations of NH₃, NH₄⁺, NO₃⁻ and HNO₃. The 134 results show that the modelled and measured NH₃, NH₄⁺ and NO₃⁻ concentrations are in good agreement, 135 with R² values above 0.89 and slopes near 1.00, indicating that the thermodynamic analysis accurately represents the aerosol state and that deviations in the calculated pH values are lower than that in modelled 136 137 NH₃(Weber et al., 2016). However, the predicted and measured concentrations of HNO₃ show a poor correlation, 138 as reported in previous studies(Ding et al., 2019; Guo et al., 2015). This may be attributed to lower gas-139 phase concentrations than particle-phase concentrations and the measurement uncertainties of HNO₃ 140 from MARGA(Rumsey et al., 2014). The development of an alternative approach is therefore required 141 to accurately represent HNO₃. 142 2.3 Sensitivity analysis and drivers of aerosol pH 143 To investigate the factors that drive changes in aerosol pH, sensitivity tests of pH to different factors, 144 including temperature, RH, SO₄²⁻, TNO₃, NH_x, Cl⁻ and NVCs, were performed. Figure S4 shows how 145 these factors affected the aerosol pH. In the sensitivity tests, we found that an elevated temperature and 146 SO₄²⁻ could decrease pH. As shown in Figure S4, particles tend to become more acidic at higher 147 temperatures. The temperature dependence of pH is mainly determined by the phase partitioning of 148 NH_3/NH_4^+ based on the equilibria $NH_3(g) \leftrightarrow NH_3(aq)$ and $NH_3(aq) + H_2O \leftrightarrow NH_4^+ + OH^-$, which are 149 governed by the temperature-dependent constants K_H and K_b, respectively(Hennigan et al., 2015; Zheng 150 et al., 2020). Figures S4 also show that elevations in NHx, NVCs and RH can increase the aerosol pH 151 and ALWC. For TNO3 and Cl-, we find their impacts on the aerosol pH were rather weak through 152 sensitivity test and thus are not discussed in detail here. 153 In this study, we also try to quantify the effects of different factors on the annual, seasonal and 154 diurnal variations of aerosol pH to identify the most important determinants. To quantify the effects of 155 individual factors on the aerosol pH, we first estimated the pH by using the ISORROPIA model with 156 initial values of the different factors, including annual, seasonal and hourly mean value of temperature, RH, SO₄²⁻, TNO₃, NH_x, Cl⁻ and NVCs. We then estimated the changed pH (pH_{change}) by varying one 157 158 factor while holding the other factors fixed. The difference in aerosol pH (Δ pH, Δ pH = pH_{change} - pH)





represents the effect of an individual factor change on the aerosol pH.

3 Results and Discussion

3.1 Long-term trends of aerosol pH

3.1.1 Trends of aerosol pH.

The 9-year time series of aerosol pH calculated by ISORROPIA II is shown in Figure 1a. A declining trend in $PM_{2.5}$ pH from 3.30 ± 0.58 in 2011 to 3.06 ± 0.55 in 2019 was observed, with the fitted decrease rate of around 0.04 unit pH per year, which may be related to chemical composition changes (Figs. S5-S6) due to the pollution control measures taken in the Yangtze River Delta (YRD) region. The Chinese government started to implement the Action Plan, a series of air pollution control policies, in September 2013, which resulted in a clear decline in $PM_{2.5}$ and its chemical components(Cheng et al., 2019; Li et al., 2019). Compared to the concentrations before the implement of the Action Plan (i.e., 2011-2012 averages), $PM_{2.5}$, SO_4^{2-} , NH_x and NVCs after the implement of the Action Plan (i.e., 2018-2019 averages) decreased by 35.8%, 37.6%, 9.6% and 81.0%, respectively, while NO_3^- increased by 1.2% (Fig. S5). In terms of the chemical profiles, SO_4^{2-} , NH_4^+ and NO_3^- remain the most abundant inorganic water-soluble ions, accounting for 83.4%–94.1% of the total ions in $PM_{2.5}$. While the proportions of NH_4^+ and NO_3^- increased continuously (increased by 2.2% and 13.1% from 2011 to 2019, respectively), those of NVCs and SO_4^{2-} decreased by 6.0% and 4.6%, respectively. Despite of the substantial change of aerosol abundance and composition, the aerosol pH shows a moderate change. The effects of chemical composition changes in $PM_{2.5}$ on the aerosol pH are further discussed in Section 3.1.2.

The PM_{2.5} in Shanghai was moderately acidic with daily pH range from 1.15 to 5.62, similar to those from other cities in China(Shi et al., 2019; Tan et al., 2018). Table S1 shows the pH data from the literature, which were calculated using thermodynamic models of different cities in summer and winter. In general, the PM_{2.5} pH ranges in Chinese cities were higher than those in US cities yet similar to those in European cities.

3.1.2 Driving factors.

Figure 1b shows the contributions of individual factors to the ΔpH from 2011 to 2019. Note that in Fig.
1b, the aerosol pH was calculated from the annual averages of input parameters. This is different from

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Sect 3.1.1, where the annual pH is the average of hourly values based on hourly observation data. The aerosol pH decreased from 3.35 in 2011 to 3.28 in 2013. The main factors that affected the pH in this period (prior to the implementation of the Action Plan) were the temperature and NVCs. Upon implementation of the Action Plan (2013-2019), the concentrations of PM2.5 and its chemical components decreased substantially (Fig. S5); hence, the role of the chemical composition in the aerosol pH become more prominent than the period of 2011-2013. The pH value continuously decreased from 3.28 in 2013 to 3.19 in 2019. Changes in SO₄²⁻ and NVCs were more important determinants of the change of aerosol pH, resulting in ΔpH of +0.38 units and -0.35 units from 2013 to 2019, respectively. Besides the effect of reduction in SO₄²⁻ (Fu et al., 2015; Xie et al., 2020), our results suggest that the change in NVCs may also play an important role in determining the trend of aerosol pH. The effects of SO₄²⁻ and NVCs on pH were much weaker during 2017-2019 than during 2013-2017, consistent with the fact that the decline in pollutant concentrations has slowed in recent years (Fig. S6). Thus, temperature and NH_x become the main drivers of the ΔpH during 2017–2019. From 2013 to 2019, changes in the NH_x and Cl⁻ were associated with 0.08 and 0.06 decreases in ΔpH , respectively, whereas TNO₃ had little effect on the ΔpH . Overall, the changes in SO₄²⁻ and NVCs were the main drivers of the ΔpH under the implemented Action Plan, and NH_x appeared to play an increasingly important role in determining the aerosol pH through the years.



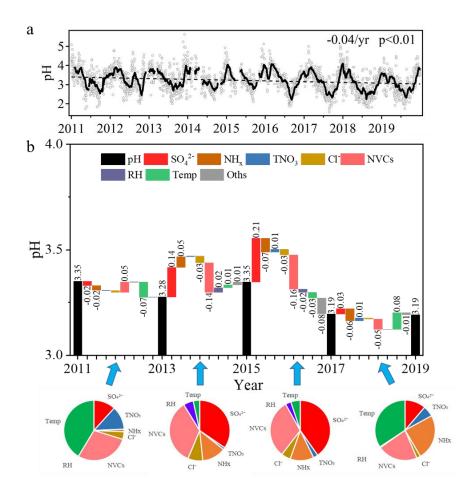


Figure 1. (a) Long-term trends in aerosol pH during 2011–2019 in Shanghai. Gray dots and black lines represent the daily pH values and 30-day moving average pH values, respectively. (b) Fractional contribution of individual drivers to the change in aerosol pH during 2011–2019. RH, relative humidity; Temp, temperature; NVCs, non-volatile cations; NH_x, total ammonia; TNO₃, total nitrate.

3.2 Seasonal variation

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Figure 2 shows the seasonal variations of aerosol pH in Shanghai. The average pH values were 3.33 ± 0.49 , 2.89 ± 0.49 , 2.99 ± 0.52 and 3.59 ± 0.57 in spring (March–May, MAM), summer (June–August, JJA), fall (September–November, SON) and winter (December–February, DJF), respectively. The highest aerosol pH was found in winter while the lowest pH was found in summer. This is with similar seasonal trend but generally lower levels than that observed in Beijing and other NCP cities (Tan et al.,





214 2018; Ding et al., 2019; Shi et al., 2019; Wang et al., 2020), due to the generally lower aerosol concentrations in YRD.

According to the multiphase buffer theory, the peak buffer pH, p K_a^* regulates the aerosol pH in a multiphase-buffered system, and temperature can largely drive the seasonal variation of aerosol pH through its impact on p K_a^* (Zheng et al., 2020). Figure 3 confirms this conclusion, and shows a dominant role of temperature in driving the seasonal variation of aerosol pH. The temperature was associated with a max Δ pH of 0.63 from fall to winter. Besides temperature, the main factors affecting aerosol pH are NH_x and SO₄²⁻ (Fig. 3), contributing 16% and 12% of the changes, respectively. Our results suggest a central role of temperature in the determination of seasonal variations in aerosol pH, consistent with the results of Tao and Murphy (Tao and Murphy, 2019) at six Canadian sites and the prediction by the multiphase buffer theory(Zheng et al., 2020). In comparison, some previous studies emphasized the importance of chemical compositions in seasonal variations (Tan et al., 2018; Ding et al., 2019), which is mainly due to the different sensitivity analysis methods applied.

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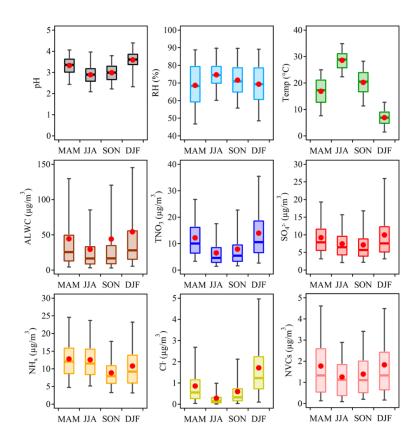


Figure 2. Seasonal patterns of the mass concentrations of major components in PM_{2.5}, relative humidity (RH), temperature (Temp), predicted aerosol liquid water content (ALWC) and aerosol pH during 2011–2019 in Shanghai.



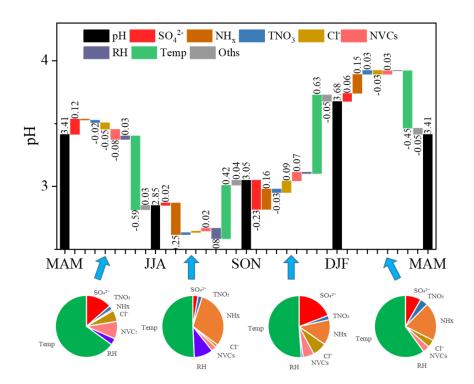


Figure 3. Fractional contributions of individual drivers to the changes in aerosol pH across the four seasons. RH, relative humidity; Temp, temperature; NVCs, non-volatile cations; NH_x , total ammonia; TNO_3 , total nitrate.

3.3 Diurnal variation

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Figure 4 shows the diurnal variations in the aerosol pH and its potential drivers. Similar to the results in Beijing(Tao et al., 2020), aerosol pH in Shanghai exhibits notable diurnal variations, being higher during nighttime.

Figure 5 shows the effects of individual factors on the diurnal variations in aerosol pH. Temperature and RH are among the main drivers of this diurnal variation of aerosol pH, with a max Δ pH of -0.22 and +0.10 units. As shown in Fig. 4, the maximum RH and ALWC occurred at approximately 5:00. After sunrise, increase of temperature resulted in an immediate drop of RH and ALWC with ALWC reached its lowest level in the afternoon. Accordingly, the minimum aerosol pH (~2.8) was also found in the afternoon with high temperature and low RH. After sunset, the decreasing temperature and increasing

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RH led to a highest aerosol pH overnight. Minor pH changes were found between 0:00 and 6:00, when temperature and RH also show minor changes. The effects of other factors on the diurnal variations in pH were notably smaller than their effects on seasonal variations, which may be attributed to the relatively small variations of chemical profiles in the course of a day. Among these chemical factors, NH_x plays the most important roles, followed by SO₄²⁻. Overall, temperature and RH are more important than the chemical compositions in controlling the diurnal variations in aerosol pH.

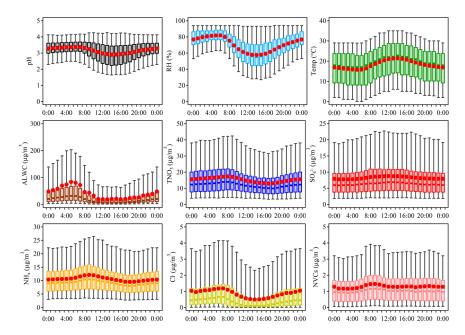


Figure 4. Diurnal patterns of the mass concentrations of major ions in PM_{2.5}, relative humidity (RH), temperature (Temp), predicted aerosol liquid water content (ALWC) and aerosol pH during 2011–2019 in Shanghai.

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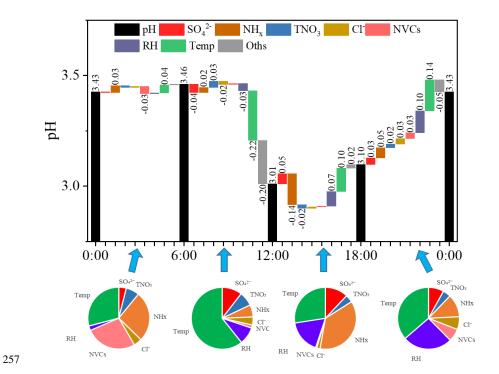


Figure 5. Fractional contributions of individual drivers to the aerosol pH difference between day and night. RH, relative humidity; Temp, temperature; NVCs, non-volatile cations; NH_x, total ammonia; TNO₃, total nitrate.

3.4 Future projections

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A series of prevention and control measures are suggested for continuous improvement of air quality, which will affect the atmospheric compositions and may subsequently affect the aerosol pH in China. To explore China's future anthropogenic emission pathways in 2015–2050, Tong et al. (Tong et al., 2020) developed a dynamic projection model, based on which different emission scenarios were created by connecting five socio-economic pathway (SSP) scenarios, five representative concentration pathways (RCP) scenarios (RCP8.5, 7.0, 6.0, 4.5 and 2.6) and three pollution control scenarios (business as usual, BAU; enhanced control policy, ECP; and best health effect, BHE). These scenarios provide a better understanding of the future trends in pollutant emissions (Tong et al., 2020).

In this study, we chose three different emission reduction scenarios as the future anthropogenic

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emission pathways, including SSP3-70-BAU, SSP2-45-ECP and SSP1-26-BHE, and tried to project the future aerosol pH levels in Shanghai. SSP1-26-BHE, which involves a combination of strong low-carbon and air pollution control policy, has the greatest emission reduction, followed by SSP2-45-ECP. SSP3-70-BAU is a reference scenario without any active actions. Figure 6 shows the emissions of SO₂, NO_x and NH₃ and predicted pH levels in China from 2015 to 2050 under the three scenarios. We also predicted the aerosol pH based on the assumption that reductions in SO₄²⁻, TNO₃ and NH_x are equivalent to reductions in their respective precursors (i.e., SO₂, NO_x and NH₃). As shown in Fig. 6, the future trend of aerosol pH changes little under the weak control policy (SSP3-70-BAU). Correspondingly, there is also little change in the predicted NO₃- partition ratio (NO₃-/ (NO₃⁻+HNO₃)). However, NH₄⁺ partition ratio (NH₄⁺/(NH₄⁺+NH₃)) increases substantially, suggesting an enhanced formation of ammonium aerosols. Under the moderate control policy (SSP2-45-ECP), the emissions of SO₂, NO_x, and NH₃ in 2050 will be reduced by 62.7%, 49.0% and 25.0%, respectively. This would result in an increase in pH of ~0.5, while the decreases in NO₃ and NH₄ partition ratios of 0.14 and 0.23 respectively. That is, more nitrate and ammonium will exist in the gas phase as HNO₃ and NH₃. With the strict control policy (SSP1-26-BHE), the emissions of SO₂, NO_x and NH₃ in 2050 will be reduced by 86.9%, 74.9% and 41.7%, respectively, and the pH will further increase to 4.30 in 2050. It is noted that the predicted pH values in the SSP1-26-BHE model decrease suddenly during 2030-2040 and then continue to rise, which is due to the decrease in ALWC and increase in pH brought by the decrease of SO₄²⁻, TNO₃ and NH_x during 2015-2030. These variations result in a transition to the HNO₃-sensitive regime during 2030-2040, where partitioning gradually shifts from aerosol-phase nitrate to the gaseous phase(Nenes et al., 2020a). Therefore, the decrease in TNO₃ during 2030–2040 lead to a notable decrease of pH. This can also be seen from Fig. 6e, where NO₃ partition ratio begins to decrease gradually and reaches the HNO3-sensitive regime and then rises again under the strict control policy, resulting in NO3partition ratio increased from 0.92 in 2015 to 1.00 in 2050. According to the strict control policy, we also found that NH₄⁺ partition ratio has dropped significantly from 0.37 in 2015 to 0.02 in 2050. Because of China's commitment in reducing CO2, we expect a further roadmap of emission control similar to the SSP1-26-BHE scenario. In this case, the corresponding aerosol pH in eastern China will increase by ~0.9, resulting in 8% more NO₃ and 35% less NH₄ partitioning/formation in the aerosol phase. This would suggest a largely reduced benefit of NH₃ and NO_x emission control in mitigating haze pollution in eastern



300 China.

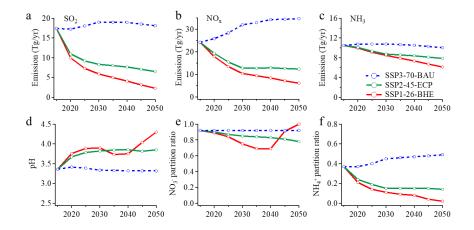


Figure 6. Emissions of SO₂ (a), NO₃ (b), NH₃ (c), predicted pH (d), NO₃ partition (NO₃ + HNO₃)) (e) and NH₄ partition (NH₄ + NH₄)) (f) in China from 2015 to 2050 under the three scenarios published in Tong et al. (Tong et al., 2020)

4 Conclusion

The aerosol pH values at an urban site in Shanghai during 2011–2019 were calculated using ISORROPIA II. The aerosol pH decreased from 3.30 ± 0.58 in 2011 to 3.06 ± 0.58 in 2019, which is related to changes in the chemical composition under the implementation of Action Plan in the YRD region. The implementation of the Action Plan leads to -35.8%, -37.6%, -9.6%, -81.0% and +1.2% changes in concentrations of $PM_{2.5}$, SO_4^{2-} , NH_x , NVCs and NO_3^{-} in YRD during this period. Then we quantified the contributions of individual factors to the difference in aerosol pH from 2011 to 2019. Before the implementation of the Action Plan (2011-2013), the main factors that affected the pH were the temperature and NVCs. During the implementation of the Action Plan (2013-2019), SO_4^{2-} and NVCs were the most important determinants of the aerosol pH and were attributed to ΔpH of +0.38 and -0.35 units, respectively. Meanwhile, changes in NH_x and CI^- were responsible for decreases in aerosol pH of 0.08 and 0.06 pH units, respectively, whereas the effect of TNO_3 was negligible. NH_x appeared to play an increasingly important role in determining aerosol pH through the years.

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Distinct seasonal variations in the aerosol pH were observed, with maximum and minimum aerosol pH of 3.59 ± 0.57 in winter (DJF) and 2.89 ± 0.49 in summer (JJA), respectively. Temperature through its impact on pKa* can largely drive the seasonal variation of aerosol pH in a multiphase-buffered system. Besides temperature, the main factors affecting the seasonal variation of aerosol pH are NH_x and SO₄²⁻, which contributed to +16% and +12% pH change, respectively. The diurnal cycle of particle pH is driven by the combined effects of temperature and relative humidity with a maximum ΔpH of -0.22 and +0.10 units, respectively. The effects of chemical factors on the diurnal variations in pH were notably smaller than the effects on seasonal variations, which may be attributed to the limited variations of chemical profiles in the course of a day. Finally, to explore the effects of China's future anthropogenic emission control pathways on aerosol pH and compositions, we chose three different emission reduction scenarios proposed by Tong et al.(Tong et al., 2020) for future haze mitigation, naming SSP3-70-BAU, SSP2-45-ECP and SSP1-26-BHE as case studies. We estimated that the future trend of aerosol pH and NO₃ partition ratio change little under the weak control policy (SSP3-70-BAU). However, NH₄⁺ partition ratio increases substantially, suggesting an enhanced formation of ammonium aerosols. The results also demonstrate that future aerosol pH will increase under both strict control policy (SSP1-26-BHE) and moderate control policy (SSP2-45-ECP), but more drastically under former scenario. The significant increase in aerosol pH with strict control policy will lead to more nitrate partitioning in aerosol phase, hence inhibiting future PM_{2.5} pollution control. This finding implicates that variations in aerosol pH will feedback to multiphase formation pathways of aerosols in the atmosphere. In other words, proportional reductions in precursors and follow-up variations in aerosol pH are evidently necessary to be taken into account for future efforts in mitigating haze pollution.

Author Contributions

HS, HW, and CH conceived and led the study. MZ conducted the field measurements and carried out the data analysis. MZ and GZ performed model simulations. MZ, HS, HW, CH, GZ, LQ, SZ, DH, YC, JA discussed the results. LQ, SZ, DH, SL, ST, QW, RY, YM, CC conducted the measurements at the station. MZ, HS and GZ wrote the manuscript with input from all co-authors.





- 346 Supplement
- 347 The supplement is available in a separate file.
- 348 Competing interests
- 349 The authors declare that they have no conflict of interest.
- 350 Data availability
- 351 The data presented in this paper are available upon request from Hang Su (h.su@mpic.de) and Cheng
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- 359 Reference
- 360 Battaglia, M. A., Douglas, S., and Hennigan, C. J.: Effect of the Urban Heat Island on Aerosol pH,
- 361 Environmental Science & Technology, 51, 13095-13103, 10.1021/acs.est.7b02786, 2017.
- 362 Cai, S., Wang, Y., Zhao, B., Wang, S., Chang, X., and Hao, J.: The impact of the "Air Pollution Prevention
- 363 and Control Action Plan" on PM2.5 concentrations in Jing-Jin-Ji region during 2012-2020, Sci Total
- 364 Environ, 580, 197-209, 10.1016/j.scitotenv.2016.11.188, 2017.
- 365 Cheng, J., Su, J., Cui, T., Li, X., Dong, X., Sun, F., Yang, Y., Tong, D., Zheng, Y., Li, Y., Li, J., Zhang,
- 366 Q., and He, K.: Dominant role of emission reduction in PM2.5 air quality improvement in Beijing during
- 367 2013–2017: a model-based decomposition analysis, Atmospheric Chemistry and Physics, 19, 6125-6146,
- 368 10.5194/acp-19-6125-2019, 2019.
- 369 Cheng, Y., Zheng, G., Wei, C., Mu, Q., Zheng, B., Wang, Z., Gao, M., Zhang, Q., He, K., Carmichael,
- 370 G., Poscjl, U., and Su, H.: Reactive nitrogen chemistry in aerosol water as a source of sulfate during haze
- events in China, Science Advance, 2016.
- 372 Clegg, S. L., Brimblecombe, P., and Wexler, A. S.: Thermodynamic Model of the System
- 373 H⁺-NH₄⁺-Na⁺-SO₄²--NO₃⁻-Cl⁻H₂O at 298.15 K, The Journal of Physical Chemistry A, 102, 2155-
- 374 2171, 10.1021/jp973043j, 1998.
- 375 Ding, J., Zhao, P., Su, J., Dong, Q., Du, X., and Zhang, Y.: Aerosol pH and its driving factors in Beijing,
- 376 Atmospheric Chemistry and Physics, 19, 7939-7954, 10.5194/acp-19-7939-2019, 2019.
- 377 Fang, T., Guo, H., Zeng, L., Verma, V., Nenes, A., and Weber, R. J.: Highly Acidic Ambient Particles,
- 378 Soluble Metals, and Oxidative Potential: A Link between Sulfate and Aerosol Toxicity, Environ Sci





- 379 Technol, 51, 2611-2620, 10.1021/acs.est.6b06151, 2017.
- 380 Fountoukis, C. and Nenes, A.: ISORROPIA II: a computationally efficient thermodynamic equilibrium
- $381 \qquad \text{model for K+--} Ca2+-Mg2+-NH+4-Na+-SO2-4-NO-3-Cl--H2O aerosols, Atmospheric Chemistry} \\$
- and Physics, 7, 4639-4659, 2007.
- Fu, X., Guo, H., Wang, X., Ding, X., He, Q., Liu, T., and Zhang, Z.: PM2.5 acidity at a background site
- 384 in the Pearl River Delta region in fall-winter of 2007-2012, J Hazard Mater, 286, 484-492,
- 385 10.1016/j.jhazmat.2015.01.022, 2015.
- 386 Guo, H., Weber, R. J., and Nenes, A.: High levels of ammonia do not raise fine particle pH sufficiently
- 387 to yield nitrogen oxide-dominated sulfate production, Sci Rep, 7, 12109, 10.1038/s41598-017-11704-0,
- 388 2017a
- 389 Guo, H., Otjes, R., Schlag, P., Kiendler-Scharr, A., Nenes, A., and Weber, R. J.: Effectiveness of ammonia
- 390 reduction on control of fine particle nitrate, Atmospheric Chemistry and Physics, 18, 12241-12256,
- 391 10.5194/acp-18-12241-2018, 2018.
- 392 Guo, H., Liu, J., Froyd, K. D., Roberts, J. M., Veres, P. R., Hayes, P. L., Jimenez, J. L., Nenes, A., and
- Weber, R. J.: Fine particle pH and gas-particle phase partitioning of inorganic species in Pasadena,
- 394 California, during the 2010 CalNex campaign, Atmospheric Chemistry and Physics, 17, 5703-5719,
- 395 10.5194/acp-17-5703-2017, 2017b.
- 396 Guo, H., Sullivan, A. P., Campuzano-Jost, P., Schroder, J. C., Lopez-Hilfiker, F. D., Dibb, J. E., Jimenez,
- 397 J. L., Thornton, J. A., Brown, S. S., Nenes, A., and Weber, R. J.: Fine particle pH and the partitioning of
- 398 nitric acid during winter in the northeastern United States, Journal of Geophysical Research:
- 399 Atmospheres, 121, 10,355-310,376, 10.1002/2016jd025311, 2016.
- 400 Guo, H., Xu, L., Bougiatioti, A., Cerully, K. M., Capps, S. L., Hite, J. R., Carlton, A. G., Lee, S. H.,
- Bergin, M. H., Ng, N. L., Nenes, A., and Weber, R. J.: Fine-particle water and pH in the southeastern
- 402 United States, Atmospheric Chemistry and Physics, 15, 5211-5228, 10.5194/acp-15-5211-2015, 2015.
- 403 He, P., Alexander, B., Geng, L., Chi, X., Fan, S., Zhan, H., Kang, H., Zheng, G., Cheng, Y., Su, H., Liu,
- 404 C., and Xie, Z.: Isotopic constraints on heterogeneous sulfate production in Beijing haze, Atmospheric
- 405 Chemistry and Physics, 18, 5515-5528, 10.5194/acp-18-5515-2018, 2018.
- 406 Hennigan, C. J., Izumi, J., Sullivan, A. P., Weber, R. J., and Nenes, A.: A critical evaluation of proxy
- 407 methods used to estimate the acidity of atmospheric particles, Atmospheric Chemistry and Physics, 15,
- 408 2775-2790, 10.5194/acp-15-2775-2015, 2015.
- Huang, X. H. H., Bian, Q., Ng, W. M., Louie, P. K. K., and Yu, J. Z.: Characterization of PM2.5 Major
- 410 Components and Source Investigation in Suburban Hong Kong: A One Year Monitoring Study, Aerosol
- 411 and Air Quality Research, 14, 237-250, 10.4209/aaqr.2013.01.0020, 2014.
- 412 Jia, S., Wang, X., Zhang, Q., Sarkar, S., Wu, L., Huang, M., Zhang, J., and Yang, L.: Technical note:
- 413 Comparison and interconversion of pH based on different standard states for aerosol acidity
- 414 characterization, Atmospheric Chemistry and Physics, 18, 11125-11133, 10.5194/acp-18-11125-2018,
- 415 2018
- 416 Li, C., Hu, Y., Chen, J., Ma, Z., Ye, X., Yang, X., Wang, L., Wang, X., and Mellouki, A.: Physiochemical
- 417 properties of carbonaceous aerosol from agricultural residue burning: Density, volatility, and
- 418 hygroscopicity, Atmospheric Environment, 140, 94-105, 10.1016/j.atmosenv.2016.05.052, 2016.
- 419 Li, H., Cheng, J., Zhang, Q., Zheng, B., Zhang, Y., Zheng, G., and He, K.: Rapid transition in winter
- 420 aerosol composition in Beijing from 2014 to 2017: response to clean air actions, Atmospheric Chemistry
- 421 and Physics, 19, 11485-11499, 10.5194/acp-19-11485-2019, 2019.
- 422 Li, W., Xu, L., Liu, X., Zhang, J., Lin, Y., Yao, X., Gao, H., Zhang, D., Chen, J., Wang, W., Harrison, R.





- 423 M., Zhang, X., Shao, L., Fu, P., Nenes, A., and Shi, Z.: Air pollution-aerosol interactions produce more
- bioavailable iron for ocean ecosystems, Science Advance, 3, e1601749, 2017.
- 425 Liu, M., Huang, X., Song, Y., Xu, T., Wang, S., Wu, Z., Hu, M., Zhang, L., Zhang, Q., Pan, Y., Liu, X.,
- 426 and Zhu, T.: Rapid SO2 emission reductions significantly increase tropospheric ammonia concentrations
- over the North China Plain, Atmospheric Chemistry and Physics, 18, 17933-17943, 10.5194/acp-18-
- 428 17933-2018, 2018.
- 429 Masiol, M., Squizzato, S., Formenton, G., Khan, M. B., Hopke, P. K., Nenes, A., Pandis, S. N., Tositti,
- 430 L., Benetello, F., Visin, F., and Pavoni, B.: Hybrid multiple-site mass closure and source apportionment
- 431 of PM2.5 and aerosol acidity at major cities in the Po Valley, Sci Total Environ, 704, 135287,
- 432 10.1016/j.scitotenv.2019.135287, 2020.
- 433 Nah, T., Guo, H., Sullivan, A. P., Chen, Y., Tanner, D. J., Nenes, A., Russell, A., Ng, N. L., Huey, L. G.,
- 434 and Weber, R. J.: Characterization of aerosol composition, aerosol acidity, and organic acid partitioning
- 435 at an agriculturally intensive rural southeastern US site, Atmospheric Chemistry and Physics, 18, 11471-
- 436 11491, 10.5194/acp-18-11471-2018, 2018.
- 437 Nenes, A., Pandis, S. N., Weber, R. J., and Russell, A.: Aerosol pH and liquid water content determine
- 438 when particulate matter is sensitive to ammonia and nitrate availability, Atmospheric Chemistry and
- 439 Physics, 20, 3249-3258, 10.5194/acp-20-3249-2020, 2020a.
- 440 Nenes, A., Pandis, S. N., Kanakidou, M., Russell, A., Song, S., Vasilakos, P., and Weber, R. J.: Aerosol
- 441 acidity and liquid water content regulate the dry deposition of inorganic reactive nitrogen, Atmospheric
- 442 Chemistry and Physics Discussion, 10.5194/acp-2020-266, 2020b.
- 443 Pye, H. O. T., Zuend, A., Fry, J. L., Isaacman-VanWertz, G., Capps, S. L., Appel, K. W., Foroutan, H.,
- 444 Xu, L., Ng, N. L., and Goldstein, A. H.: Coupling of organic and inorganic aerosol systems and the effect
- on gas-particle partitioning in the southeastern US, Atmos Chem Phys, 18, 357-370, 10.5194/acp-18-
- 446 357-2018, 2018.
- 447 Pye, H. O. T., Nenes, A., Alexander, B., Ault, A. P., Barth, M. C., Clegg, S. L., Collett Jr, J. L., Fahey, K.
- 448 M., Hennigan, C. J., Herrmann, H., Kanakidou, M., Kelly, J. T., Ku, I. T., McNeill, V. F., Riemer, N.,
- 449 Schaefer, T., Shi, G., Tilgner, A., Walker, J. T., Wang, T., Weber, R., Xing, J., Zaveri, R. A., and Zuend,
- 450 A.: The acidity of atmospheric particles and clouds, Atmospheric Chemistry and Physics, 20, 4809-4888,
- 451 10.5194/acp-20-4809-2020, 2020.
- 452 Qiao, L., Cai, J., Wang, H., Wang, W., Zhou, M., Lou, S., Chen, R., Dai, H., Chen, C., and Kan, H.:
- 453 PM2.5 constituents and hospital emergency-room visits in Shanghai, China, Environ Sci Technol, 48,
- 454 10406-10414, 10.1021/es501305k, 2014.
- Rumsey, I. C., Cowen, K. A., Walker, J. T., Kelly, T. J., Hanft, E. A., Mishoe, K., Rogers, C., Proost, R.,
- 456 Beachley, G. M., Lear, G., Frelink, T., and Otjes, R. P.: An assessment of the performance of the Monitor
- 457 for AeRosols and GAses in ambient air (MARGA): a semi-continuous method for soluble compounds,
- 458 Atmospheric Chemistry and Physics, 14, 5639-5658, 10.5194/acp-14-5639-2014, 2014.
- 459 Shi, X., Nenes, A., Xiao, Z., Song, S., Yu, H., Shi, G., Zhao, Q., Chen, K., Feng, Y., and Russell, A. G.:
- 460 High-Resolution Data Sets Unravel the Effects of Sources and Meteorological Conditions on Nitrate and
- Its Gas-Particle Partitioning, Environ Sci Technol, 53, 3048-3057, 10.1021/acs.est.8b06524, 2019.
- Song, S., Gao, M., Xu, W., Shao, J., Shi, G., Wang, S., Wang, Y., Sun, Y., and McElroy, M. B.: Fine-
- 463 particle pH for Beijing winter haze as inferred from different thermodynamic equilibrium models,
- 464 Atmospheric Chemistry and Physics, 18, 7423-7438, 10.5194/acp-18-7423-2018, 2018.
- 465 Su, H., Cheng, Y., and Poschl, U.: New Multiphase Chemical Processes Influencing Atmospheric
- 466 Aerosols, Air Quality, and Climate in the Anthropocene, Acc Chem Res, 53, 2034-2043,





- 467 10.1021/acs.accounts.0c00246, 2020.
- 468 Tan, T., Hu, M., Li, M., Guo, Q., Wu, Y., Fang, X., Gu, F., Wang, Y., and Wu, Z.: New insight into PM2.5
- 469 pollution patterns in Beijing based on one-year measurement of chemical compositions, Sci Total
- 470 Environ, 621, 734-743, 10.1016/j.scitotenv.2017.11.208, 2018.
- 471 Tao, W., Su, H., Zheng, G., Wang, J., Wei, C., Liu, L., Ma, N., Li, M., Zhang, Q., Pöschl, U., and Cheng,
- 472 Y.: Aerosol pH and chemical regimes of sulfate formation in aerosol water during winter haze in the
- 473 North China Plain, Atmospheric Chemistry and Physics, 20, 11729-11746, 10.5194/acp-20-11729-2020,
- 474 2020.
- 475 Tao, Y. and Murphy, J. G.: The sensitivity of PM2.5 acidity to meteorological parameters and chemical
- 476 composition changes: 10-year records from six Canadian monitoring sites, Atmos. Chem. Phys., 19,
- 477 9309-9320, 10.5194/acp-19-9309-2019, 2019.
- 478 Tong, D., Cheng, J., Liu, Y., Yu, S., Yan, L., Hong, C., Qin, Y., Zhao, H., Zheng, Y., Geng, G., Li, M.,
- 479 Liu, F., Zhang, Y., Zheng, B., Leon, C., and Zhang, Q.: Dynamic projection of anthropogenic emissions
- 480 in China: methodology and 2015-2050 emission pathways under a range of socio-economic, climate
- 481 policy, and pollution control scenarios, Atmospheric Chemistry and Physics, 20, 5729-5757,
- 482 10.5194/acp-20-5729-2020, 2020.
- 483 Wang, H., Ding, J., Xu, J., Wen, J., Han, J., Wang, K., Shi, G., Feng, Y., Ivey, C. E., Wang, Y., Nenes, A.,
- 484 Zhao, Q., and Russell, A. G.: Aerosols in an arid environment: The role of aerosol water content,
- 485 particulate acidity, precursors, and relative humidity on secondary inorganic aerosols, Sci Total Environ,
- 486 646, 564-572, 10.1016/j.scitotenv.2018.07.321, 2019.
- 487 Wang, S., Wang, L., Li, Y., Wang, C., Wang, W., Yin, S., and Zhang, R.: Effect of ammonia on fine-
- 488 particle pH in agricultural regions of China: comparison between urban and rural sites, Atmospheric
- 489 Chemistry and Physics, 20, 2719-2734, 10.5194/acp-20-2719-2020, 2020.
- Weber, R. J., Guo, H., Russell, A. G., and Nenes, A.: High aerosol acidity despite declining atmospheric
- 491 sulfate concentrations over the past 15 years, Nature Geoscience, 9, 282-285, 10.1038/ngeo2665, 2016.
- 492 Xie, Y., Wang, G., Wang, X., Chen, J., Chen, Y., Tang, G., Wang, L., Ge, S., Xue, G., Wang, Y., and Gao,
- 493 J.: Nitrate-dominated PM2.5 and elevation of particle pH observed in urban Beijing during the winter of
- 494 2017, Atmospheric Chemistry and Physics, 20, 5019-5033, 10.5194/acp-20-5019-2020, 2020.
- 495 Zheng, B., Tong, D., Li, M., Liu, F., Hong, C., Geng, G., Li, H., Li, X., Peng, L., Qi, J., Yan, L., Zhang,
- 496 Y., Zhao, H., Zheng, Y., He, K., and Zhang, Q.: Trends in China's anthropogenic emissions since 2010 as
- 497 the consequence of clean air actions, Atmospheric Chemistry and Physics, 18, 14095-14111,
- 498 10.5194/acp-18-14095-2018, 2018.
- 499 Zheng, G., Su, H., Wang, S., Andreae, M. O., Poschl, U., and Cheng, Y.: Multiphase buffer theory
- explains contrasts in atmospheric aerosol acidity, Science 369, 1374-1377, 2020.
- 501 Zhou, M., Qiao, L., Zhu, S., Li, L., Lou, S., Wang, H., Wang, Q., Tao, S., Huang, C., and Chen, C.:
- 502 Chemical characteristics of fine particles and their impact on visibility impairment in Shanghai based on
- 503 a 1-year period observation, J Environ Sci, 48, 151-160, 10.1016/j.jes.2016.01.022, 2016.