



Long-term Trends in PM_{2.5} Chemical Composition and Its Impact on Aerosol Properties: Field Observations from 2007 to 2020 in Pearl River Delta, South China

Yunfeng He^{1,4}, Xiang Ding^{1,2,3}, Quanfu He⁸, Yuqing Zhang⁹, Duohong Chen¹⁰, Tao Zhang¹⁰,
Kong Yang^{1,4}, Junqi Wang^{1,4}, Qian Cheng^{1,4}, Hao Jiang^{1,4}, Zirui Wang^{1,4}, Ping Liu^{1,4},
Xinming Wang^{1,2,3}, and Michael Boy^{5,6,7}

¹State Key Laboratory of Advanced Environmental Technology, Guangzhou Institute of Geochemistry, Chinese Academy of Sciences, Guangzhou 510640, China

²Guangdong–Hong Kong–Macao Joint Laboratory for Environmental Pollution and Control, Guangzhou Institute of Geochemistry, Chinese Academy of Science, Guangzhou 510640, China

³Guangdong Key Laboratory of Environmental Protection and Resources Utilization, Guangzhou Institute of Geochemistry, Guangzhou 510640, China

⁴College of Earth and Planetary Sciences, University of Chinese Academy of Sciences, Beijing 100049, China

⁵Institute for Atmospheric and Earth Systems Research, University of Helsinki, Helsinki 00014, Finland

⁶Atmospheric Modelling Centre – Lahti, Lahti University Campus, Lahti 15140, Finland

⁷School of Engineering Sciences, Lappeenranta-Lahti University of Technology, Lappeenranta 53850, Finland

⁸Thrust of Earth, Ocean and Atmospheric Sciences, The Hong Kong University of Science and Technology (Guangzhou), Guangzhou 511453, China

⁹School of Environment and Safety Engineering, North University of China, Taiyuan 030051, China

¹⁰State Environmental Protection Key Laboratory of Regional Air Quality Monitoring, Guangdong Environmental Monitoring Center, Guangzhou 510308, China

Correspondence: Xiang Ding (xiangding@gig.ac.cn) and Michael Boy (michael.boy@helsinki.fi)

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Abstract. Long-term data on PM_{2.5} chemical composition provide essential information for evaluating the effectiveness of air pollution control measures and understanding the evolving mechanisms of secondary species formation in the real atmosphere. This study presented field measurements of PM_{2.5} and its chemical composition at a regional background site in the Pearl River Delta (PRD) from 2007 to 2020. PM_{2.5} concentration declined significantly from 87.1 ± 15.5 to $34.0 \pm 11.3 \mu\text{g m}^{-3}$ ($-4.0 \mu\text{g m}^{-3} \text{ yr}^{-1}$). The proportion of secondary species increased from 57 % to 73 % with the improvement in air quality. Among these species, sulfate (SO_4^{2-}) showed a sharp decline, while nitrate (NO_3^-) exhibited a moderate decrease. Consequently, the proportion of NO_3^- in 2020 doubled relative to 2007. In addition, we further found that SO_4^{2-} reduction ($-10 \% \text{ yr}^{-1}$) lagged behind SO_2 reduction ($-13 \% \text{ yr}^{-1}$), while NO_3^- reduction ($-6 \% \text{ yr}^{-1}$) outpaced that of NO_2 ($-3 \% \text{ yr}^{-1}$). These contrasting trends were associated with an increase in sulfur oxidation rate (SOR) and a decrease in nitrogen oxidation rate (NOR). Changes in PM_{2.5} chemical composition also influenced aerosol physicochemical properties, such as aerosol pH (0.04 yr^{-1}), aerosol liquid water content (ALWC, $-1.1 \mu\text{g m}^{-3} \text{ yr}^{-1}$), and the light extinction coefficient ($-21.44 \text{ Mm}^{-1} \text{ yr}^{-1}$). Given important roles of aerosol acidity and ALWC in the heterogeneous reactions, these changes may further inhibit the formation of secondary species in the atmosphere, particularly secondary organic aerosols.

1 Introduction

Particulate matter with aerodynamic diameter less than 2.5 μm (PM_{2.5}) is a major air pollutant with significant implications for global climate, air quality, and human health (Burnett et al., 2018; Chen et al., 2021; Ding et al., 2021; Pye et al., 2021; Vohra et al., 2022). PM_{2.5} comprises a complex mixture of primary and secondary components. Primary components, including primary organic aerosol (POA), elemental carbon (EC), and metal ions (e.g., K⁺, Ca²⁺, Na⁺, Mg²⁺), are mostly emitted from anthropogenic activities. Secondary components, such as secondary organic aerosols (SOA) and secondary inorganic aerosols (SIA; i.e., sulfate, nitrate, and ammonium), are formed through oxidation of gaseous pollutants (SO₂, NO_x, and VOCs, etc.) and partition processes. China experienced rapid economic growth and urbanization in the past decades. To address severe air pollution, the Chinese government issued the Air Pollution Prevention and Control Action Plan in 2013 (Geng et al., 2024). As a result, the chemical composition in PM_{2.5} over China changed significantly (Geng et al., 2019). This change has an important impact on aerosol acidity, ALWC, and light extinction (Nguyen et al., 2016; Pye et al., 2020; Liu et al., 2022).

Acidity, defined as pH, is a crucial aerosol property that affects human health, ecosystems and climate (Nenes et al., 2020; Su et al., 2020; Song et al., 2024). Low pH increases solubility of metals associated with mineral dust (Fang et al., 2017). Previous epidemiological studies revealed that exposure to acidic PM_{2.5} is relevant to high mortality and morbidity (Gwynn et al., 2000; Zhang et al., 2022a). Additionally, aerosol acidity and ALWC regulate the gas–particle partitioning of semi-volatile compounds, as well as chemical reaction rates in the atmosphere, highlighting their importance for the atmospheric lifetime of pollutants (Pye et al., 2020; Nenes et al., 2021). Aqueous uptake is an important formation pathway for secondary species (Yu et al., 2005; Kawamura and Bikkina, 2016; Liu et al., 2021). As an abundant medium, ALWC can enhance their formation (Carlton and Turpin, 2013; Zheng et al., 2015). By modifying particle ability to be activated into cloud condensation nuclei (CCN), ALWC can further influence the climate system (Duan et al., 2019). Furthermore, Attwood et al. (2014) reported that the changes in ALWC significantly influenced the aerosol light extinction and radiative forcing. Therefore, it is necessary to explore the trends of pH and ALWC under the changes in PM_{2.5} chemical composition.

Aerosol pH and ALWC are determined by the presence of acidic components (i.e., sulfate and nitrate), alkaline components (i.e., ammonium) (Seinfeld et al., 1998), and meteorological conditions, such as temperature and relative humidity (Wang et al., 2022a). However, aerosol acidity is more sensitive to dominant chemical species rather than meteorological conditions (Wu et al., 2023). According to sensitivity tests,

T-H₂SO₄ (SO₂ + SO₄²⁻) and T-NH₃ (NH₃ + NH₄⁺) have the most dominant negative and positive contributions to pH variation, respectively (Wu et al., 2023). The aerosol pH exhibited noticeable spatial heterogeneity. For example, the pH values in North China (e.g., Beijing (3.0–4.9), Zhengzhou (4.5), Anyang (4.8)) (Liu et al., 2017; Wang et al., 2020) were generally higher than those in South China (e.g., Guangzhou (−0.04–0.81), Shanghai (3.06–3.30), South China Sea (1.7)) (Fu et al., 2015; Wang et al., 2022a; Zhou et al., 2022). This could be attributed to the higher fraction of sulfate observed in southern China (Geng et al., 2017; Liu et al., 2023) and the elevated ammonia emissions in northern China (Liu et al., 2023). Zhou et al. (2022) reported a downward trend of pH (from 3.33 to 3.06) at a rate of -0.24 yr^{-1} in the Yangtze River Delta (YRD) from 2011 to 2019. Conversely, Fu et al. (2015) reported an upward trend of pH (from −0.30 to 0.81) in the PRD during 2007–2012. However, this study did not cover the post-2013 period, a key period for air quality improvement.

Chemical composition in PM_{2.5} also affects atmospheric visibility through light scattering and absorption. Light scattering is dominated by hydrophilic components, such as organic mass (OM), (NH₄)₂SO₄, and NH₄NO₃, while light absorption is largely driven by light-absorbing carbon (Wang et al., 2012). To estimate the light extinction coefficient (b_{ext}), the first Interagency Monitoring of Protected Visual Environments (IMPROVE) equation was developed by the U.S. National Park Service with support from the U.S. Environmental Protection Agency (EPA) (Malm et al., 1994; EPA, 2003), but this equation tended to underestimate (overestimate) the highest (lowest) b_{ext} values. Consequently, the revised IMPROVE equation was then proposed (Malm and Hand, 2007; Pitchford et al., 2007). However, the scattering/absorbing efficiency (MSE/MAE) in the revised equation is an approximation based on measurements from clean areas. In addition, the calculation of hygroscopic growth factor ($f(\text{RH})$) in the revised equation depends on relative humidity (RH) and particle size distribution (or aerosols mass), but does not account for the chemical composition in aerosols, which has been shown to significantly affect $f(\text{RH})$ (Li et al., 2021). These simplifications could lead to large discrepancies in polluted regions. For instance, the deviations between observed and estimated b_{ext} values were reported as 15 %, 36 %, and 37 % in Xi'an, Shanghai, and Guangzhou, respectively (Jung et al., 2009; Cao et al., 2012; Cheng et al., 2015). Thus, region-specific adjustments are necessary to reflect the impact of particle composition on these parameters from site to site.

Many long-term monitoring programs have been implemented to formulate pollution control strategies and explore underlying factors of aerosol properties variation. For example, the IMPROVE program in the United States, initiated in 1985, tracks visibility trends and their driving factors (Hand et al., 2024). The Southeastern Aerosol Research and Char-

acterization (SEARCH) network, established in 1998, provides detailed insights into aerosol chemistry and precursor gases (Blanchard et al., 2013). In Europe, the European Monitoring and Evaluation Program (EMEP) has operated since 2004 to address air pollution issues related to acidification, eutrophication, and climate impacts (EMEP, 2024). In China, long-term PM_{2.5} monitoring began in Hong Kong, where the Environmental Protection Department (HKEPD) initiated comprehensive chemical composition measurements in 1999. Subsequent collaborations expanded the monitoring network to encompass the Guangdong–Hong Kong–Macao Greater Bay Area in 2015 (HKEPD, 2021; Chow et al., 2022).

As one of the most outstanding areas for air pollution improvement, the PRD region first met National Ambient Air Quality Standards (AQS) for annual average PM_{2.5} ($35\text{ }\mu\text{g m}^{-3}$) in 2015 (Department of Ecology and Environment of Guangdong Province, 2016). PM_{2.5} in the PRD ($32.2\text{--}46.1\text{ }\mu\text{g m}^{-3}$) was significantly lower than the YRD ($44.8\text{--}67.1\text{ }\mu\text{g m}^{-3}$), the North China Plain (NCP) ($64.0\text{--}101.9\text{ }\mu\text{g m}^{-3}$), and other regions ($45.1\text{--}65.4\text{ }\mu\text{g m}^{-3}$) (Zhang et al., 2019). Regarding chemical composition, nitrate was the dominant species in the YRD and the NCP. However, OM constitutes the largest fraction in the PRD, similar to other low PM_{2.5} areas worldwide (Zhang et al., 2007; Ming et al., 2017; Geng et al., 2019; Wang et al., 2022b; Yang et al., 2023). So far, long-term studies of PM_{2.5} chemical components in the PRD remain scarce. Fu et al. (2014) observed substantial reductions in organic carbon (OC) and SO_4^{2-} , but relatively stable NO_3^- during 2007–2011, alongside increased aerosol pH and decreased light extinction (Fu et al., 2014, 2015, 2016). However, this study did not cover the post-2013 period, a key period for air quality improvement. Yan et al. (2020) conducted a meta-analysis (2004–2019), identifying three stages of decline, rise, and stabilization in the fractions of secondary species. Chow et al. (2022) reported the reduction in NO_3^- (66 %), EC (60 %), hopanes (75 %), and K^+ (60 %) exceeding that of PM_{2.5} (40 %), confirming effective control of vehicle emissions and biomass burning in Hong Kong (2008–2017). In addition, these studies observed an unproportional relationships between $\text{SO}_4^{2-}/\text{SO}_2$, as well as $\text{NO}_3^-/\text{NO}_2$, but the underlying reasons remain unclear. Furthermore, long-term trends of aerosol acidity, ALWC, and light extinction, which are highly dependent on PM_{2.5} chemical composition, were not fully investigated. These limitations underscore the need for a more comprehensive, long-term study to explore the underlying mechanisms behind these changes in the PRD.

Our study presents a comprehensive analysis of 532 quartz filter-based PM_{2.5} samples collected over 14 years (2007–2020, wintertime) at a regional background station. We examined the evolving PM_{2.5} chemical composition, focusing on both primary and secondary species. The unproportional relationships between $\text{SO}_4^{2-}/\text{SO}_2$, as well as $\text{NO}_3^-/\text{NO}_2$ will be discussed. In addition, we also analyzed variations of

aerosol pH, ALWC, and light extinction under the influence of changes in PM_{2.5} chemical composition.

2 Methodology

2.1 Field sampling

The typical Asian monsoon climate influences the PRD region. In summertime, prevailing southwesterly winds bring humid and clean air mass from South China Sea or the northwestern Pacific Ocean. In contrast, during the fall and winter, prevailing northerly winds carry dry and polluted air mass from northern continent. Additionally, the region is often influenced by high-pressure ridges in the fall and winter, which results in a low boundary layer and a high frequency of inversion. These conditions facilitate the accumulation of pollutants. Consequently, PM_{2.5} and other pollutants show a distinct summer–winter contrast, with significantly elevated pollutant levels in fall and winter (Fu et al., 2014; Chow et al., 2022). Our field campaigns were mainly conducted from October to December.

The sampling site, Wanqingsha (WQS; 22.42°N , 113.32°E), is located in a rural area of Guangzhou and the center of the PRD region (Fig. 1). Local anthropogenic emissions have limited influence on this site due to low traffic flow and few factories in the surrounding area. This makes it an ideal background site for investigating regional air pollution. Twenty-four-hour sampling was conducted using a PM_{2.5} sampler equipped with quartz filters ($20.3\text{ cm} \times 25.4\text{ cm}$) at a flow rate of $1.1\text{ m}^3\text{ min}^{-1}$. The quartz filters were pre-baked at 450°C for four hours prior to field sampling and stored at -20°C after sampling until analysis. Blank samples were collected during the sampling periods. A total of 532 samples were collected and analyzed in this study. Gaseous pollutants data (SO_2 , NO_2 , NO , and O_3) and meteorological parameters were obtained from an air quality monitoring station operated at WQS. The gaseous pollutant data during 2012–2013 were unavailable because the station was under maintenance.

2.2 Chemical analysis

Because the determination of OC and EC are highly sensitive to analytical conditions, different thermal-optical methods may lead to discrepancies in the OC/EC measurements (Khan et al., 2012; Giannoni et al., 2016). To minimize potential measurement biases and enhance data comparability, we used the same thermal-optical carbon analyzer and followed the same analytical protocol throughout the study. Specifically, the OC and EC were determined by the thermal-optical transmittance (TOT) method (NIOSH, 1999) using an OC/EC analyzer (Sunset Laboratory Inc., USA), with a punch ($1.5 \times 1.0\text{ cm}$) of the sampled filters. The samples were analyzed by stepwise heating. First, the sample was heated sequentially to 870°C (310°C for 60 s, 475°C for

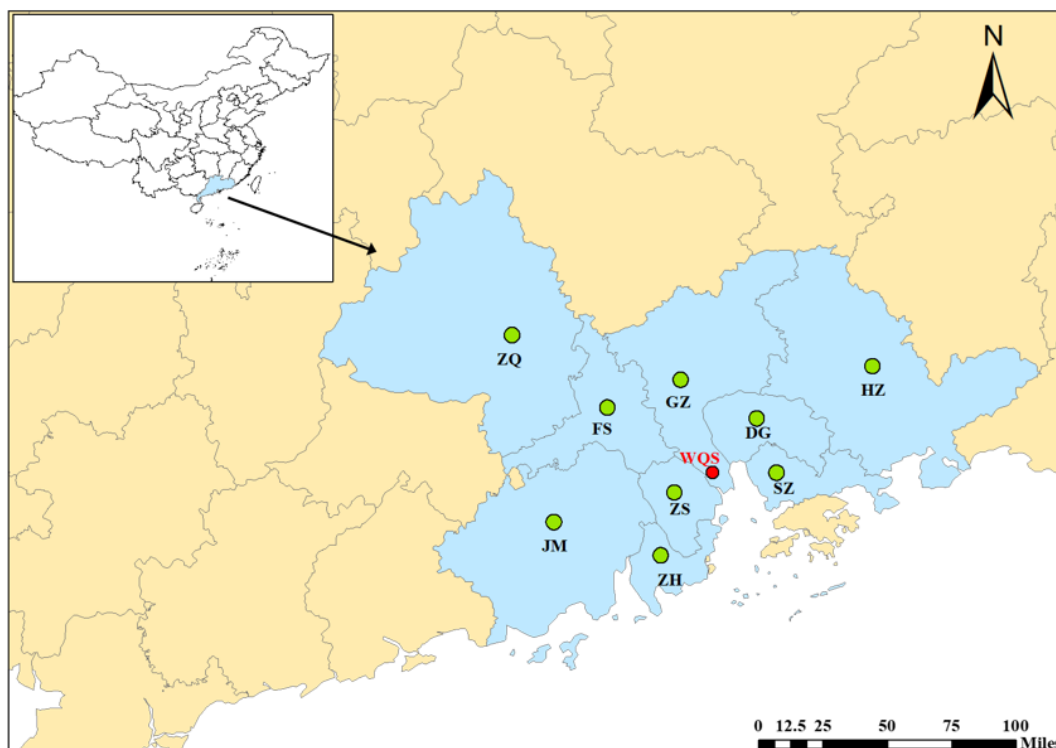


Figure 1. The PRD region consists of Guangzhou (GZ), Shenzhen (SZ), Zhuhai (ZH), Dongguan (DG), Foshan (FS), Huizhou (HZ), Zhongshan (ZS), Zhaoqing (ZQ), and Jiangmen (JM). The sampling site WQS is located in the central area of this region.

60 s, 620 °C for 60 s, and 870 °C for 90 s) under a pure helium (He) atmosphere. OC was volatilized and a portion of it underwent pyrolysis, forming pyrolyzed carbon during this period. After cooling, the sample was reheated under a 2 % O₂ / He atmosphere up to 920 °C (625 °C for 30 s, 700 °C for 30 s, 775 °C for 30 s, 850 °C for 30 s, and 920 °C for 30 s) to oxidize EC and pyrolyzed carbon.

For the water-soluble inorganic ions, a punch (5.06 cm²) of the filters was extracted twice with 10 mL ultrapure Milli-Q water (18.2 MΩ cm per 25 °C) each for 15 min using an ultrasonic ice-water bath. The total water extracts (20 mL) were filtered through a 0.22 μm pore size filter and then stored in a pre-cleaned HDPE bottle. The cations (i.e., Na⁺, NH₄⁺, K⁺, Mg²⁺, and Ca²⁺) and anions (i.e., Cl⁻, NO₃⁻, and SO₄²⁻) were analyzed with an ion-chromatography system (Metrohm, 883 Basic IC plus). Due to the negative mass artifacts associated with the volatilization of ammonium nitrate, the measured concentrations of NO₃⁻ and NH₄⁺ may be underestimated (Chow et al., 2005; Yu et al., 2006). Cations were measured using a Metrohm Metrosep C4–100 column with 2 mmol L⁻¹ sulfuric acid as the eluent. Anions were measured using a Metrohm Metrosep A sup 5–150 column equipped with a suppressor. The anion eluent was a solution of 3.2 mmol L⁻¹ Na₂CO₃ and 1.0 mmol L⁻¹ NaHCO₃.

2.3 Quality assurance/quality control (QA/QC)

Prior to OC / EC analysis, we calibrated the instrument using glucose standards at multiple concentrations. The instrument responses were highly linear ($R^2 > 0.99$) and the relative errors between measured and prepared concentrations were within ±5 %. The method detection limits (MDLs) were 0.01–0.05 μg m⁻³ for the OC, EC, cations, and anions. Ion balance was employed as a quality control check in the anion/cation analysis. A significant linear correlation ($R^2 = 0.97$) was observed between anions and cations, with a slope of 0.82 for all PM_{2.5} samples. This slope, being close to unity, indicated that all the significant ions were resolved.

Blank samples were analyzed in the same way as field samples. All the OC / EC and cation/anion data were corrected by subtracting the field blank samples. As shown in Fig. S1 in the Supplement, all measured components exhibited minimal variability in the blank filter samples. This indicated that the influence of analytical and sampling biases related to blank subtraction and experimental procedures was limited, further supporting the reliability of the long-term trends observed in this study.

Before data analysis, all data were manually inspected and outliers (i.e., $X_{75\%} + 3(X_{75\%} - X_{25\%})$) were removed to rule out the influence of extreme concentrations on overall trends. The change rates were calculated using the slopes derived from Theil–Sen regression and evaluated for statistical sig-

nificance via the non-parametric Mann–Kendall test, providing a robust and reliable assessment of temporal variations.

2.4 Estimation of primary organic carbon (POC) and secondary organic carbon (SOC)

EC is a product of carbon fuel-based combustion processes and is exclusively associated with primary emissions, whereas OC can be formed from both direct emissions and secondary pathways. Differentiation between POC and SOC is indispensable for probing atmospheric aging processes of organic aerosols, but all available methods for estimating POC and SOC are highly uncertain. The EC–tracer method was widely used (Turpin and Huntzicker, 1991, 1995). Given that EC is emitted exclusively from primary combustion sources (e.g., fossil fuel and biomass burning), it is commonly used as a tracer for POC. Under this assumption, POC is estimated by multiplying EC by a representative primary OC/EC ratio, and SOC is determined as the residual between total OC and estimated POC (Eqs. 1–2). One of the most commonly used approaches to determine (OC/EC)_{pri} is the minimum OC/EC ratio approach (Castro et al., 1999), which assumes that the lowest observed OC/EC value corresponds to conditions dominated by primary emissions with negligible SOC formation. In addition, Pio et al. (2011) recommended using the 5% percentile of observed OC/EC values instead, and Wu and Yu (2016) proposed minimum R squared (MRS) method to obtain (OC/EC)_{pri}.

$$\text{POC} = \left(\frac{\text{OC}}{\text{EC}} \right)_{\text{pri}} \times \text{EC} \quad (1)$$

$$\text{SOC} = \text{OC} - \text{POC} \quad (2)$$

However, a previous study revealed that EC–tracer method relied on the fixed (OC/EC)_{pri} and tended to overestimate SOC (Kim et al., 2012). Recently, Liao et al. (2023) proposed Bayesian Inference (BI) approach and suggested it more accurately estimated POC and SOC compared to the conventional method. The BI approach adopts a probabilistic framework that combines prior knowledge (in the form of prior distributions of the *K* values) with observational data (OC, EC, and SIA) to estimate the source contributions to OC. The BI model assumes that observed OC is a linear combination of contributions from POC and SOC tracers (e.g., EC and SO₄^{2−}), and it uses the Markov Chain Monte Carlo (MCMC) technique to derive posterior distributions for the *K* values. This treatment allows the approach to update parameter estimates based on actual measurements and offers more flexibility. In this study, we used BI approach which has the convenience of relying only on the commonly available mass concentrations of EC and SO₄^{2−} to estimate POC and SOC. They can be calculated as following:

$$\text{SOC} = K_{\text{EC}} \times \text{EC} + K_{\text{SO}_4^{2-}} \times \text{SO}_4^{2-} \quad (3)$$

$$\text{POC} = \text{OC} - \text{SOC} \quad (4)$$

Where *K*_{EC} and *K*_{SO₄^{2−}} are parameters calculated by BI approach in R language, the details can be found in previous research (Liao et al., 2023). The *K* values represent the proportionality constants that link OC to its respective tracers (e.g., *K*_{EC} for EC-to-POC and *K*_{SO₄^{2−}} for sulfate-to-SOC). The variations of *K* value are shown in Fig. S2.

Because SOC represents only the carbonaceous portion of the organic aerosol, while SOA includes the entire mass of organic compounds formed through secondary processes, including associated non-carbon atoms (e.g., hydrogen, oxygen, nitrogen). To better evaluate the atmospheric mass loading of organic aerosols and their mass proportions in PM_{2.5}, it is important to convert SOC to SOA. Given intense photochemical reactions and larger fractions of aged aerosols in the PRD, a higher conversion factor of 2.4 was employed to convert SOC to SOA (Yan et al., 2020).

2.5 Prediction of aerosol acidity and ALWC

The thermodynamics model ISORROPIA-II has been widely used to calculate aerosol pH (Nenes et al., 1998; Fountoukis and Nenes, 2007; Wen et al., 2018; Zhou et al., 2022). Meteorological parameter (temperature and relative humidity), aerosol phase water-soluble ions (SO₄^{2−}, NO₃[−], Cl[−], NH₄⁺, Na⁺, K⁺, Mg²⁺, Ca²⁺) and gaseous precursors (HNO₃, HCl, NH₃) are needed for the model execution. Given high relative humidity (RH) in the PRD region, the model was run by assuming aerosol solutions were metastable (only a liquid phase) in forward mode; previous studies suggested this would produce better performance than the stable state solution (solid + liquid) (Guo et al., 2015; Bougiatioti et al., 2016). A recent study suggested that the model was run in the forward mode but did not include gas-phase data, which may capture the general trend of aerosol acidity but underestimate pH (Fang et al., 2025).

We used the data collected in Guangzhou Institute of Geochemistry (GIG) to run the model, and the result (G1) showed there was a great agreement in gaseous ammonia between simulation and observation (Fig. S3). Due to the lack of gas-phase concentrations during our campaign period in WQS, we ran the model by performing iterations to avoid the underestimation of pH. Only aerosol phase data were used as input for the first run. Then we added the output gas-phase concentrations from the first run to the initial aerosol-phase chemical concentrations to derive total (gas and aerosol phase) concentrations, which serve as input for the second run and so on. To determine an optimal number of iterations, we ran GIG data (without gas-phase data) to compare with G1. Our result indicated four iterations generated the closest outcome to G1 when gas-phase data was unavailable (Fig. S4).

3 Results and discussion

3.1 Long-term trends of PM_{2.5} and its chemical composition in wintertime

In response to severe particle pollution, the Chinese government issued the Air Pollution Prevention and Control Action Plan in 2013. Due to the strengthened emission controls in the PRD, primary pollutants have been reduced significantly over the past decades (Bian et al., 2019). Based on daily PM_{2.5} concentrations in our samples, we divided the data into five groups according to interim targets recommended by the World Health Organization (WHO) in 2021 (World Health Organization, 2021): IT0 (PM_{2.5} > 75 µg m⁻³), IT1 (75 µg m⁻³ > PM_{2.5} > 50 µg m⁻³), IT2 (50 µg m⁻³ > PM_{2.5} > 37.5 µg m⁻³), IT3 (37.5 µg m⁻³ > PM_{2.5} > 25 µg m⁻³), and IT4 (25 µg m⁻³ > PM_{2.5}). The majority of samples fell into T0 and T1 categories (41 %–100 %) before 2013, while this ratio quickly decreased (8 %–60 %) during 2013–2020 (Fig. S5), indicating successful implementation of air pollution mitigation strategies after 2013.

Annual average concentrations of PM_{2.5} and its chemical composition are presented in Fig. 2a and Table S1. The variations in mass fractions of these components are shown in Fig. S6. It is worth noting that the year 2020 was characterized by unprecedented emission reductions associated with COVID-19 lockdowns (Wang et al., 2021), which may have temporarily affected the trends in PM_{2.5} and its chemical composition. As shown in Table S2, a sensitivity analysis was conducted to evaluate the uncertainty introduced by including 2020 in the long-term trends analysis. The results suggested that this anomalous year would not introduce large bias on the overall long-term trends. From 2007 to 2020, PM_{2.5} concentrations exhibited a significant decline from 87.1 ± 15.5 to 34.0 ± 11.3 µg m⁻³, at a rate of -4.0 µg m⁻³ yr⁻¹ ($p < 0.01$). This trend aligns with the previous results from meta-analysis (-3.9 µg m⁻³ yr⁻¹) and regional simulation (-4.0 µg m⁻³ yr⁻¹) in the PRD (Zhang et al., 2019; Yan et al., 2020), affirming WQS can serve as a regional background site. During this period of air quality improvement, OM (defined as OC × 1.6) (Yang et al., 2023) remained the most abundant component (25 %–47 %) in PM_{2.5}, followed by SO₄²⁻ (16 %–26 %), NO₃⁻ (7 %–18 %), NH₄⁺ (7 %–10 %) and EC (3 %–8 %). Other ions, such as Cl⁻, Na⁺, K⁺, Mg²⁺, and Ca²⁺, contributed less than 3 % each to the mass of PM_{2.5} (Fig. S6). In China, desulfurization regulation for power plants was enforced around 2005, resulting in a notable decline (-7 %) in the proportion of SO₄²⁻ (Fig. 2b–c). However, the installation of denitrification devices on power plants began in late 2011 and started to take effect in 2012. With the delayed implementation of NO_x emissions control measures compared to those for SO₂ (Fu et al., 2014; Reuter et al., 2014; Geng et al., 2017; Qu et al., 2017), the mass fractions of NO₃⁻ increased by 10 %. By the

end of 2020, the proportion of NO₃⁻ had doubled compared to 2007, approaching the level of SO₄²⁻. The proportion of OM also increased from 35.5 % to 42.7 %. Consequently, future air pollution control efforts need to focus on reducing OM and NO₃⁻ concentrations to continue improving air quality in the PRD region.

The chemical composition of PM_{2.5} can be categorized into primary species and secondary species. Primary species consist of POA, EC, and metal ions (e.g., Cl⁻, Na⁺, K⁺, Mg²⁺, Ca²⁺ etc.). Secondary species include SOA and SIA (SO₄²⁻, NO₃⁻, and NH₄⁺). Our results indicated that secondary species consistently dominated over primary species in PM_{2.5} composition, accounting for 54 % to 79 % of the total mass (Fig. 3a). Additionally, secondary species declined at a faster rate (-2.45 µg m⁻³ yr⁻¹, $p < 0.01$) compared to primary species (-1.48 µg m⁻³ yr⁻¹, $p < 0.01$), indicating reduction of secondary species had more contribution in particulate pollution mitigation. Although Yan et al. (2020) also observed a decline trend in concentrations of secondary species after 2008 in the PRD, the proportion of secondary species was stable around 80 %, which was higher than our study. This might stem from the fact that the EC-tracer method applied by the previous study would overestimate SOC, which made secondary species increase (Kim et al., 2012). Our result showed that the average concentration of SOC estimated by the BI approach was ~ 30 % lower than that by the EC-tracer method. In addition, the correlation between oxalic acid (a typical secondary organic molecular marker) and SOC estimated by the BI approach ($r = 0.62$, $p < 0.05$) was stronger than that with SOC estimated by the EC-tracer method ($r = 0.54$, $p < 0.05$), indicating the estimation from BI approach was more reliable (Fig. S7). We calculated K values on an annual basis to estimate SOC, whereas Yan et al. (2020) used a fixed value of (OC / EC)_{pri} (the minimum value of all collected data) to estimate SOC. As a result, the proportion of secondary species in this study showed greater variability than that of Yan et al. (2020). In addition, we analyzed the proportion of secondary species under different pollution levels. Figure 3b showed that the mass fraction of secondary species increased significantly from 57 % to 73 % with the improvement of air quality (from IT0 to IT4). This meant secondary species play an increasingly prominent role under lower pollution conditions.

3.2 Annual variations of primary species and secondary species in wintertime

3.2.1 Primary species

The trends of individual components in PM_{2.5} can be seen in Figs. S8–S9. POA exhibited the most significant decline at a rate of -0.97 µg m⁻³ yr⁻¹ (-9 % yr⁻¹, $p < 0.01$). EC serves as a tracer for primary combustion, K⁺ serves as a tracer for biomass burning, and Ca²⁺ could be applied to track dust-related sources (Turpin and Huntz-

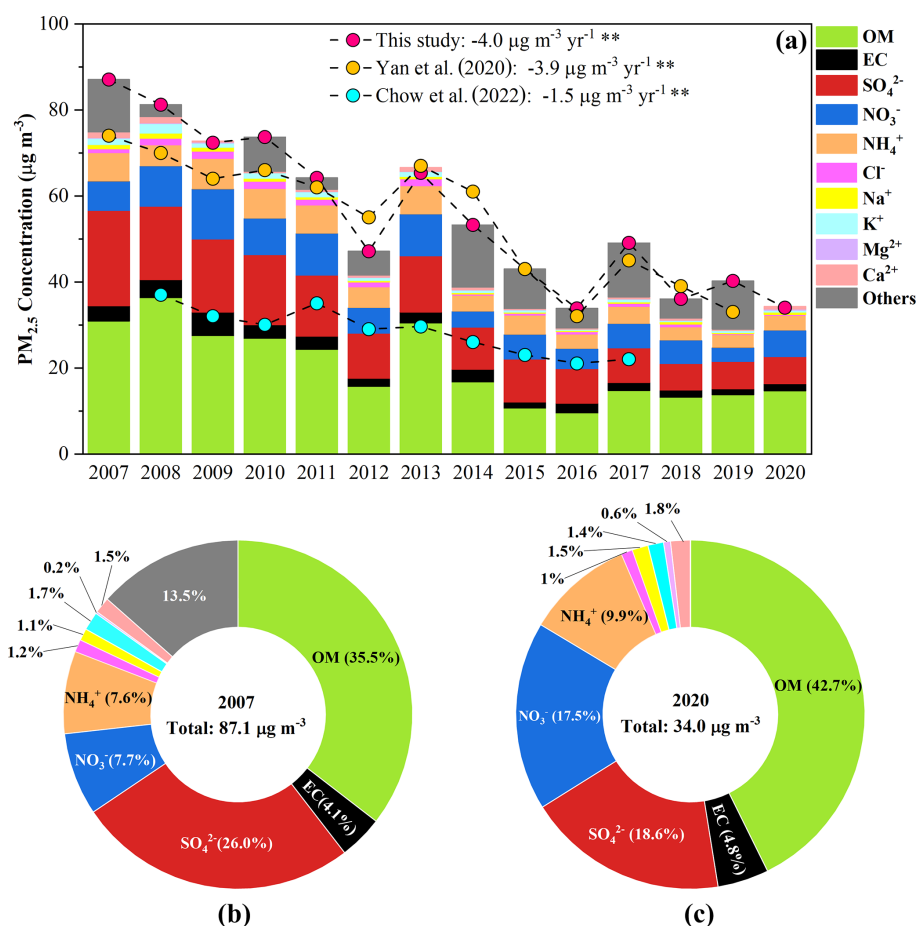


Figure 2. (a) Trends of PM_{2.5} and its major components. Bars represent concentrations of chemical compositions and circles represent concentrations of PM_{2.5} in different studies. (b, c) The comparison of the mass fractions of major components in 2007 and 2022.

icker, 1991; Zhu et al., 2017). Our result showed that the relative reductions in EC ($-9\% \text{ yr}^{-1}$), K⁺ ($-12\% \text{ yr}^{-1}$) and Ca²⁺ ($-11\% \text{ yr}^{-1}$) were greater than that of PM_{2.5} ($-7\% \text{ yr}^{-1}$), confirming that control measures for these sources had been effective. Cl⁻ and Na⁺ also showed decline trends at rates of $-0.10 \mu\text{g m}^{-3} \text{ yr}^{-1}$ ($-10\% \text{ yr}^{-1}$), $-0.05 \mu\text{g m}^{-3} \text{ yr}^{-1}$ ($-9\% \text{ yr}^{-1}$), respectively ($p < 0.01$). Marine emission is considered the biggest source of Cl⁻ in fine particle. However, anthropogenic sources such as coal combustion and biomass burning also had non-negligible impacts on it (Luo et al., 2019). After excluding the influence of anthropogenic sources (Section S1 and Fig. S10), Cl⁻ showed only a slight decline ($-2\% \text{ yr}^{-1}$), suggesting that the contribution from marine emissions to PM_{2.5} remained stable in general.

3.2.2 Secondary species

SO₄²⁻ showed a clear decrease at a rate of $-1.13 \mu\text{g m}^{-3} \text{ yr}^{-1}$ ($-10\% \text{ yr}^{-1}$, $p < 0.01$), whereas NO₃⁻ and NH₄⁺ showed moderate declines ($-0.40 \mu\text{g m}^{-3} \text{ yr}^{-1}$, $-6\% \text{ yr}^{-1}$; $-0.31 \mu\text{g m}^{-3} \text{ yr}^{-1}$, $-6\% \text{ yr}^{-1}$, respectively, $p < 0.05$).

Previous studies reported that the volatilization of ammonium nitrate during sampling can cause negative mass artifacts, leading to the underestimation of both NO₃⁻ (8%–16%) (Chow et al., 2005) and NH₄⁺ (10%–28%) (Yu et al., 2006). The volatilization is highly dependent on relative humidity and temperature. However, such losses are expected to be systematic over time and therefore are unlikely to significantly affect the general trends in this study, because our measurements were conducted in the same season. Strong correlations between SO₄²⁻/SO₂, as well as between NO₃⁻/NO₂ were observed (Fig. S11), suggesting that reductions of SO₄²⁻ and NO₃⁻ were mainly driven by their gaseous precursors. With RH rising, the slopes of SO₄²⁻/SO₂, as well as NO₃⁻/NO₂, increased, indicating enhanced conversion of primary pollutants to secondary species. The generally lower intercepts observed in the NO₃⁻/NO₂ regression compared to those in the SO₄²⁻/SO₂ regression can be explained by the semi-volatile nature of nitrate (Yu et al., 2006). The formation of HNO₃, gaseous precursor of NO₃⁻, is suppressed under very low NO₂ level. Therefore, the reaction in R2 tends to proceed to

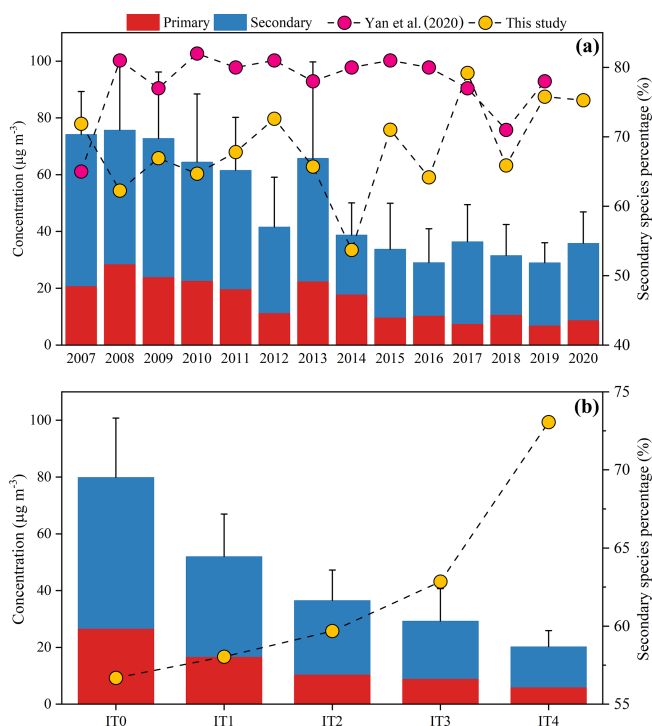


Figure 3. The variations in primary and secondary species during 2007–2020 (a) and their variations under different pollutant levels (b). Bars represent concentrations of them and circles represent the mass proportion of secondary species in PM_{2.5}. Secondary species (account for 54 %–79 %) dominated over primary species. The proportion of secondary species increased from 57 % to 73 % with improvement of air quality (From IT0 to IT4). The error bars represent the standard deviation of the total concentration, calculated as the sum of primary and secondary species.

the left. This facilitates partitioning of NO₃⁻ into gas phase, leading to less accumulation of NO₃⁻ in particle phase. In contrast, sulfate is the least volatile among all the inorganic aerosol components (Kang et al., 2022), allowing it to be stably retained in the particle phase once formed. Notably, SO₄²⁻ reduction (−10 % yr⁻¹) lagged behind SO₂ reduction (−13 % yr⁻¹), while NO₃⁻ reduction (−6 % yr⁻¹) outpaced that of NO₂ (−3 % yr⁻¹) (Fig. 4). Other studies have also observed these disproportionate changes, but the reasons behind remained unclear (Blanchard et al., 2013; Geng et al., 2019; Yan et al., 2020; Chow et al., 2022).

Here, we calculated SOR and NOR (Li et al., 2023) described in Eqs. (5)–(6), where n refers to the molar concentration. The higher SOR and NOR represent more efficient conversion of gaseous species into secondary aerosols.

$$\text{SOR} = \frac{n[\text{SO}_4^{2-}]}{n[\text{SO}_4^{2-}] + n[\text{SO}_2]} \quad (5)$$

$$\text{NOR} = \frac{n[\text{NO}_3^-]}{n[\text{NO}_3^-] + n[\text{NO}_2]} \quad (6)$$

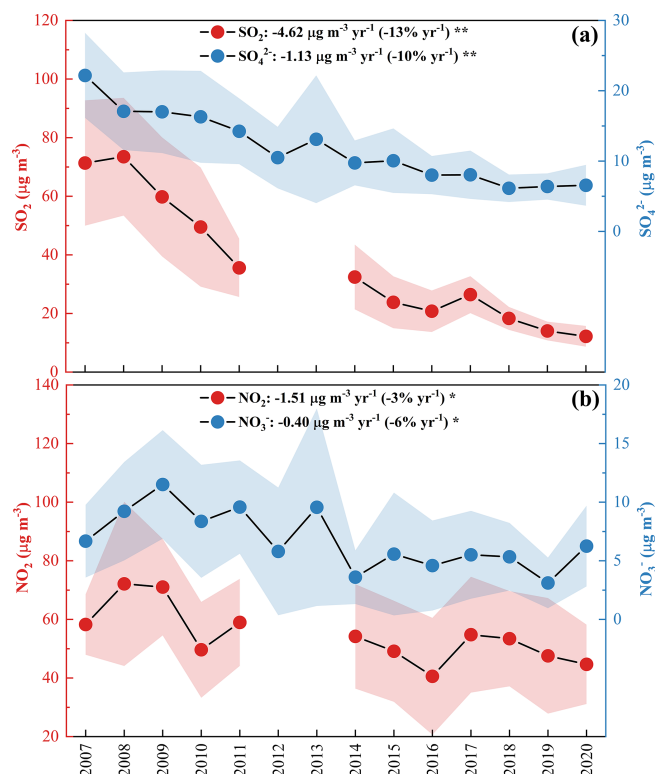


Figure 4. (a) Annual variations in SO₄²⁻ and SO₂. (b) Annual variations in NO₃⁻ and NO₂. The shaded region indicates the uncertainty bounds. One asterisk, two asterisks denote p value < 0.05, 0.01, respectively. In the PRD, SO₄²⁻ and SO₂ showed a more significant decline than NO₃⁻ and NO₂. The reduction of SO₄²⁻ (−10 % yr⁻¹) was slower than SO₂ (−13 % yr⁻¹), while reduction of NO₃⁻ (−6 % yr⁻¹) was faster than NO₂ (−3 % yr⁻¹).

As shown in Fig. 5a, the SOR value in 2020 (0.26 ± 0.09) was 44 % higher than that in 2007 (0.18 ± 0.07). In general, SOR exhibited a significant upward trend during 2007–2020, increasing at a rate of 0.005 yr^{-1} ($3 \% \text{ yr}^{-1}$, $p < 0.05$). The more efficient SO₄²⁻ formation from SO₂ oxidation slowed down the reduction of SO₄²⁻ alongside decreasing SO₂ levels. Gas-phase oxidation of SO₂ followed by neutralization and aerosol phase condensation, is an important SO₄²⁻ formation pathway (Berndt et al., 2023). Heterogeneous processes, including SO₂ transfer to the aerosol phase, dissolution, and oxidation by oxidants such as H₂O₂ and O₃, also contribute significantly in polluted regions (Wang et al., 2016; Liu et al., 2021). The solubility and effective Henry's law constant of SO₂ are positively pH-dependent (Seinfeld et al., 1998). Higher pH promotes the dissolution of SO₂ in water, which will enhance SO₄²⁻ formation. ALWC plays a key role in determining the aqueous oxidation rate and mass transfer. In addition, high temperature can facilitate both gas-phase and aqueous-phase reactions. As shown in Fig. S12, there were strong positive correlations between SOR and O₃ ($r = 0.45$),

temperature ($r = 0.48$), as well as ALWC ($r = 0.23$). But there was no significant correlation between SOR and pH. A possible explanation is that hydrogen ions facilitate aqueous-phase oxidation of SO₂ by H₂O₂, which will offset the effect of reduced SO₂ solubility under low pH conditions (Liu et al., 2021). To assess the impacts of these factors on SOR and eliminate dimensional and order of magnitude effects, a normalized multiple linear regression was developed as below:

$$\text{SOR} = 0.025 \times \text{O}_3 + 0.017 \times \text{ALWC} + 0.019 \times \text{Temperature} + 0.190 \quad (7)$$

The prediction of SOR showed good agreement with the observations (Fig. S13a). The larger regression coefficients of O₃ and temperature, along with their stronger correlations with SOR, suggested that the increase of SOR was mainly driven by the two factors. Our result showed that ALWC exhibited a downward trend (Fig. 6b), which exerted a negative influence on SOR. Although O₃ concentration did not show an obvious trend at our measurement station (Fig. S14a), a previous study suggested that there was a rapid increase of O₃ in the PRD after 2013 (Cao et al., 2024). Meanwhile, the temperature also rose slightly ($p < 0.05$) during the past decade (Fig. S14b). Consequently, a significant upward trend was observed in SOR.

NOR did not show a clear temporal trend. However, the average NOR values before 2013 ($\text{NOR} > 0.07$) were significantly higher than those in subsequent years ($\text{NOR} < 0.07$) (Fig. 5b, t -test, $p < 0.01$), which explained the greater reduction in NO₃[−] compared to NO₂. NO₃[−] can be formed during both daytime and nighttime. During the day, HNO₃ is produced via the gas–phase reaction between OH and NO₂ and then neutralized by NH₃ to produce NO₃[−] (Reactions R1–R2) (Calvert and Stockwell, 1983). During nighttime, NO₂ can also be oxidized by O₃ to generate NO₃ which further reacts with NO₂ to produce N₂O₅. Heterogeneous uptake of N₂O₅ is a vital nitrate formation pathway during nighttime (Reactions R3–R5) (Finlayson-Pitts et al., 1989). Figure S15 showed that there were positive correlations between NOR and O₃ ($r = 0.12$), pH ($r = 0.25$), as well as ALWC ($r = 0.55$). Different from SOR, higher temperature prevents N₂O₅ formation and promotes evaporation of HNO₃ from particle phase to gas phase, resulting a negative correlation between NOR and temperature ($r = -0.18$). The result of normalized multiple linear regression for NOR is as below:

$$\text{NOR} = 0.012 \times \text{O}_3 + 0.025 \times \text{ALWC} - 0.011 \times \text{Temperature} + 0.008 \times \text{pH} + 0.081 \quad (8)$$

The prediction of NOR showed good agreement with the observations (Fig. S13b). The largest regression coefficient and the strongest correlation between ALWC and NOR suggested that the change in NOR was primarily driven by ALWC. The lower ALWC levels after 2013 (Fig. 6b) suppressed heterogeneous formation pathway of nitrate and enhanced the partitioning of nitrate from particle phase into gas

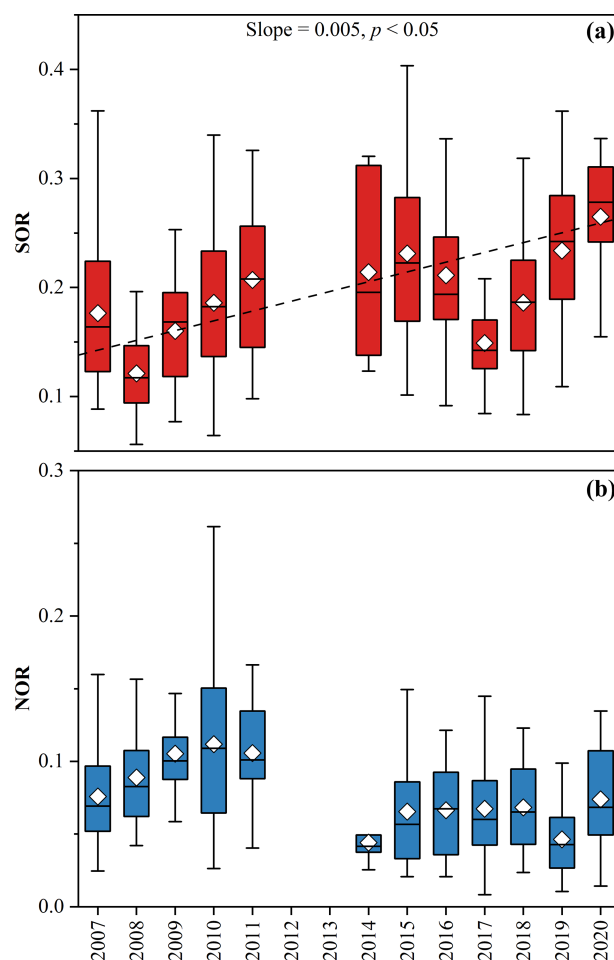
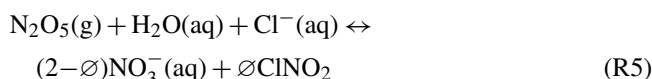
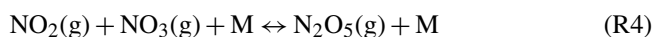
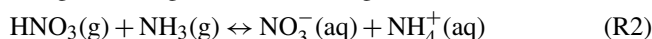


Figure 5. The variations in SOR (a) and NOR (b). A dramatic increase in SOR was observed (0.005 yr^{-1} , $p < 0.05$) and the SOR value in 2020 (0.26 ± 0.09) was 44 % higher than that in 2007 (0.18 ± 0.07). Although there was no significant trend in NOR, the values before 2012 were higher than those after 2013.

phase, leading to the overall lower NOR after 2013.



The gas–particle conversion of NH₃ could be affected by anions in particle–phase. Due to the decrease in $2 \times n(\text{SO}_4^{2-}) + n(\text{NO}_3^-)$ (Fig. S16), less NH₃ was needed to neutralize H₂SO₄ and HNO₃. This resulted in a slight decline in NH₄⁺ ($-0.31 \mu\text{g m}^{-3} \text{ yr}^{-1}$, $p < 0.01$), while NH₃ emissions remained steady (Geng et al., 2019).

SOA is formed through the oxidation of VOCs, followed by partitioning from gas phase to particle phase. Al-

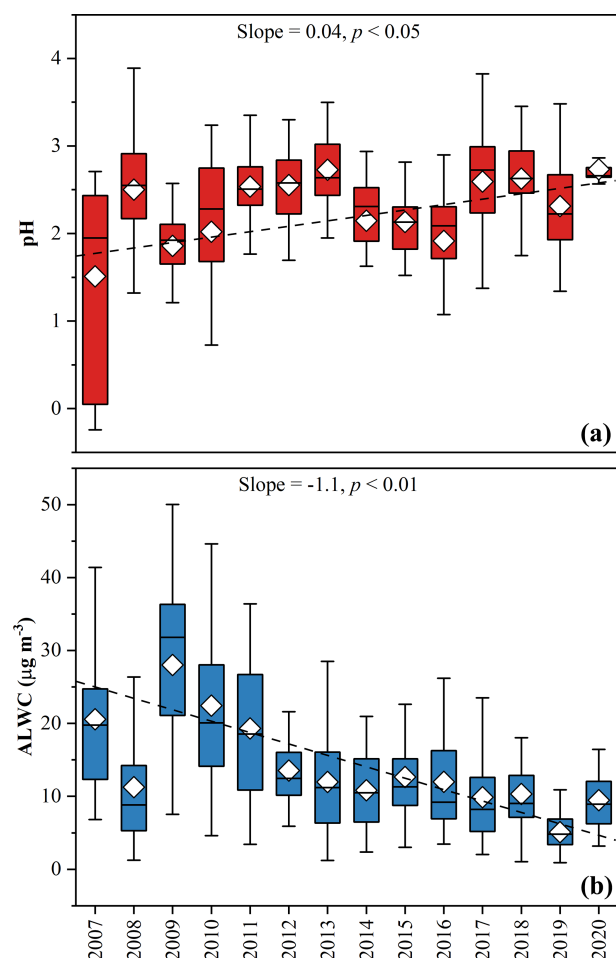


Figure 6. (a) Annual average pH increased at a rate of 0.04 yr^{-1} . (b) Annual average ALWC decreased at a rate of $-1.1 \mu\text{g m}^{-3} \text{ yr}^{-1}$.

though VOCs emission kept rising (Bian et al., 2019; Guo et al., 2024) and concentration of O₃ fluctuated during the past decade, SOA declined significantly ($-0.74 \mu\text{g m}^{-3} \text{ yr}^{-1}$, $p < 0.01$). Previous studies have demonstrated that ALWC is a key factor driving the partitioning of organic compounds from the gas phase into the particle phase, thereby promoting SOA formation (Ervens et al., 2011; Carlton and Turpin, 2013; Attwood et al., 2014). Nguyen et al. (2015) observed concurrent decreasing trends in ALWC and OC in the Southeast U.S., and further suggested that anthropogenic inorganic species modulated SOA formation through ALWC effects. In addition, higher aerosol acidity has been shown to enhance SOA formation via acid-catalyzed reactions (Surratt et al., 2007). These findings indicated that the reduction in SOA during our study period aerosol could be attributed to the changes in acidity and ALWC, which will be discussed in Sect. 3.3. As SOA accounted for more than 50 % of OM, more efforts are needed to reduce it.

3.3 Impact of changes of major components on aerosol pH and ALWC

ISORROPIA II, a thermodynamic equilibrium model for the $\text{K}^+ - \text{Ca}^{2+} - \text{Mg}^{2+} - \text{NH}_4^+ - \text{Na}^+ - \text{SO}_4^{2-} - \text{NO}_3^- - \text{Cl}^- - \text{H}_2\text{O}$ aerosol system (Fountoukis and Nenes, 2007), has been widely applied to estimate aerosol pH and ALWC. Here, we applied ISORROPIA-II to estimate aerosol pH and ALWC from 2007 to 2020. Due to the unavailability of gas-phase concentrations of HNO₃, HCl, and NH₃ during our campaign period, we conducted four iterations to produce the result optimally (Sect. 2.5). The annual variations in pH and ALWC are shown in Fig. 6. As discussed earlier, the reductions in acidic components (SO_4^{2-} and NO_3^-) were greater than that in the alkaline component (NH_4^+), leading to a significant decrease in acidity. Aerosol pH increased from 1.51 ± 1.07 to 2.86 ± 0.49 , at a rate of 0.04 yr^{-1} ($p < 0.05$). A sharp increase of aerosol pH occurred during 2007–2012 due to rapid decline of SO_4^{2-} during this period. As Fu et al. (2015) did not include the gas-phase data in pH calculation, the pH values reported by them (-1.11 – 0.81) were significantly lower than those in this study (1.51 – 2.60) during the same period. The rapid reduction in hygroscopic components, especially SO_4^{2-} , led the decline of ALWC (Attwood et al., 2014). Our results showed that ALWC decreased from 20.6 ± 10.0 to $9.5 \pm 3.9 \mu\text{g m}^{-3}$, at a rate of $-1.1 \mu\text{g m}^{-3} \text{ yr}^{-1}$ ($p < 0.01$). Unexpectedly, low ALWC was observed in 2008 when SIA concentrations, which enhance the hygroscopicity of particulate matter, were at very high levels. It might be associated with low RH (Table S1). To eliminate the influence of changes in meteorological conditions, we used annual average temperature and RH during the entire campaign period as input to recalculate pH and ALWC. The results showed a clear enhancement of ALWC in 2008 (Fig. S17), and the upward trend in pH and downward trend in ALWC still persisted. This demonstrated the long-term trends of pH and ALWC were mainly driven by the changes in chemical composition of PM_{2.5}. As we discussed in Sect. 3.2.2, ALWC exhibited positive correlations with SOR and NOR. This indicated a positive feedback mechanism in which the reductions in hygroscopic components (e.g., sulfate and nitrate) led to lower ALWC, thereby suppressing SIA formation. Many studies have demonstrated that high aerosol acidity, ALWC, and O₃ will facilitate SOA formation (Ervens et al., 2011; Carlton and Turpin, 2013; Nguyen et al., 2015; Zhang et al., 2022b; Ma et al., 2024; Zhang et al., 2024). Our results also showed that SOA was positively correlated with ALWC and O₃, but negatively correlated with pH (Fig. S18). In this study, SOA declined significantly while the emission of VOCs (Bian et al., 2019; Guo et al., 2024) and O₃ (Cao et al., 2024) kept rising in the PRD. Consequently, the trend of SOA was mainly driven by the reductions in aerosol acidity and ALWC.

We further investigated changes in pH and ALWC under different pollution levels. As Fig. S19 presented, low pH occurred under elevated pollutant level (IT0), while pH values were close under other levels (IT1–IT4). Additionally, there was no significant difference between the recalculated pH (eliminating the influence of meteorological factors) and the original one, indicating limited impacts of variations of temperature and RH on aerosol acidity. ALWC showed a clear decline pattern with decreasing pollutant levels (IT0–IT4). A significant difference (18 %–35 %, $p < 0.01$) was observed between the recalculation and original results under IT2–IT4, indicating temperature and RH also exerted a significant influence on ALWC under lower pollution levels.

3.4 Impact of changes in major components on extinction coefficient

We adopted MSE/MAE suggested by Fu et al. (2016), as well as relationship between chemical composition and $f(\text{RH})$ suggested by Li et al. (2021), to reconstruct b_{ext} (herein called local parameter scheme) using Eqs. (9)–(12). [AS], [AN], [SS], [LAC] refer to mass concentrations ($\mu\text{g m}^{-3}$) of NH_4SO_4 , NH_4NO_3 , sea salt, and EC, respectively. The details of PM_{2.5} reconstruction method followed the previous study (Chow et al., 2015), i.e., $\text{AS} = 1.375 \times \text{SO}_4^{2-}$, $\text{AN} = 1.29 \times \text{NO}_3^-$, $\text{LAC} = \text{EC}$ and $\text{SS} = 1.8 \times \text{Cl}^-$. RH_{ref} was threshold of high RH, 40 % was used here.

$$b_{\text{ext}} = 6.5 \times [\text{OM}] + 2.6 \times f(\text{RH}) \times [\text{AS}] \\ + 2.4 \times f(\text{RH}) \times [\text{AN}] + 7.3 \times f(\text{RH})_{\text{ss}} \times [\text{SS}] \\ + 7.7 \times [\text{LAC}] \quad (9)$$

$$f(\text{RH}) = [(1 - \text{RH}) / (1 - \text{RH}_{\text{ref}})]^{-\gamma} \quad (10)$$

$$\gamma = 0.48 \times F + 0.59 \quad (11)$$

$$F = (\text{OC} + \text{EC}) / (\text{OC} + \text{EC} + \text{SO}_4^{2-} + \text{NO}_3^- + \text{NH}_4^+) \quad (12)$$

Our results showed that b_{ext} in the PRD decreased significantly at a rate of $-21.44 \text{ Mm}^{-1} \text{ yr}^{-1}$ ($p < 0.01$), aligning with the overall decline of PM_{2.5} (Fig. 7). However, the highest b_{ext} was unexpectedly observed in 2009, even though PM_{2.5} concentration was lower than 2007 and 2008. This anomaly could result from the highest RH in 2009. As illustrated in Fig. S20, OM dominated b_{ext} (44 %–61 %), followed by $(\text{NH}_4)_2\text{SO}_4$ (15 %–28 %) and NH_4NO_3 (6 %–13 %). The proportion of SIA (F) fluctuated during 2007–2020, which led to changes in $f(\text{RH})$ and then further influenced b_{ext} . As a result, we found that the chemical budget of b_{ext} from $(\text{NH}_4)_2\text{SO}_4$ did not exhibit a continuous decline trend, while the mass concentration and proportion of SO_4^{2-} in PM_{2.5} decreased significantly.

We also calculated b_{ext} by the revised IMPROVE equation proposed in 2007 (Malm and Hand, 2007; Pitchford et al., 2007), and compared to the local parameter scheme (Fig. S21). Generally, b_{ext} estimated by the revised

IMPROVE equation ($335.72 \pm 219.64 \text{ Mm}^{-1}$) was significantly higher than that estimated by local parameter scheme ($262.67 \pm 143.82 \text{ Mm}^{-1}$). We further investigated this discrepancy under different pollution levels. With the improvement in air quality, the difference between the two schemes narrowed gradually ($p < 0.01$). Our results indicated that the revised IMPROVE equation tended to generate higher b_{ext} in elevated pollution periods. Thus, more site-specific parameters and local parameter scheme are needed in those polluted areas to predict b_{ext} more accurately.

4 Conclusions

In this study, we conducted field measurements of PM_{2.5} mass concentrations and its chemical composition at the PRD regional background site during 2007–2020. PM_{2.5} levels showed a significant decline from 87.1 ± 15.5 to $34.0 \pm 11.3 \mu\text{g m}^{-3}$ at a rate of $-4.0 \mu\text{g m}^{-3} \text{ yr}^{-1}$. Secondary species (54 %–79 %) dominated over primary species, although the proportion was lower than that reported in a previous study (~ 80 %) in the PRD. This discrepancy could be attributed to an overestimation of SOC caused by the EC-tracer method employed in the previous study. In addition, the mass fraction of secondary species increased with the improvement in air quality, suggesting greater attention should be given to them under cleaner conditions. Among primary species, POA, EC, K^+ and Ca^{2+} exhibited significant declines. This indicated that control measures for combustion emissions, biomass burning and dust-related sources were effective. SIA displayed rapid downward trends among secondary species, particularly for SO_4^{2-} . Due to the delayed control of NO_x emissions compared to SO_2 , mass fractions of SO_4^{2-} decreased from 26.0 % to 18.6 % while NO_3^- increased from 7.7 % to 17.5 %. By the end of 2020, the proportion of NO_3^- had doubled compared to 2007, approaching the level of SO_4^{2-} . Although many previous studies have observed the disproportionate changes in $\text{SO}_4^{2-} / \text{SO}_2$ and $\text{NO}_3^- / \text{NO}_2$, underlying causes remained unclear. In this study, we found the disproportionate reductions in SO_4^{2-} ($-10 \text{ \% yr}^{-1}$) compared to SO_2 ($-13 \text{ \% yr}^{-1}$), and in NO_3^- ($-6 \text{ \% yr}^{-1}$) compared to NO_2 ($-3 \text{ \% yr}^{-1}$), which were attributed to an increase in SOR and a decrease in NOR, respectively. Correlation analysis indicated that SOR was primarily influenced by O_3 and temperature, whereas NOR was driven by ALWC.

Aerosol pH and ALWC were estimated using ISORROPIA-II. Due to the unavailability of gas-phase concentrations of HNO_3 , HCl , and NH_3 during our campaign period, we proposed a reliable approach involving four iterative calculations to obtain optimal results. Our results showed that aerosol pH increased from 1.51 ± 1.07 to 2.86 ± 0.49 at a rate of 0.04 yr^{-1} . Consistent with previous studies, we found that the impacts of meteorological factors on aerosol pH were limited, while the changes in PM_{2.5}

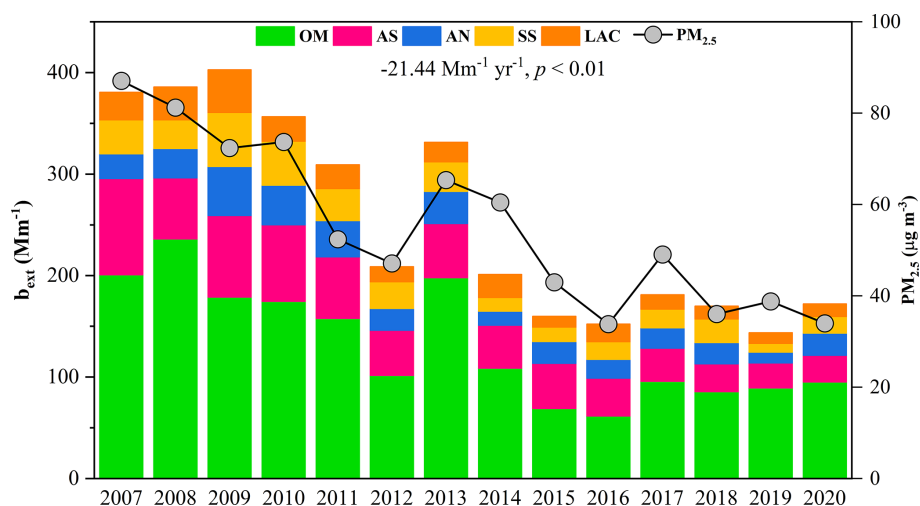


Figure 7. Light extinction coefficient (b_{ext}) in WQS during 2007–2020. It declined at a rate of $-21.44 \text{ Mm}^{-1} \text{ yr}^{-1}$ significantly ($p < 0.01$) and its trend aligned with that of PM_{2.5}.

components significantly influence aerosol pH. ALWC decreased significantly at a rate of $-1.1 \mu\text{g m}^{-3} \text{ yr}^{-1}$ and showed a clear decline pattern with decreasing pollutant levels. This might indicate presence of a positive feedback mechanism between ALWC and hygroscopic components. Given the critical roles of acidity and ALWC in the formation of secondary species, the reductions in acidity and ALWC caused by changes in PM_{2.5} major components may also suppress the formation of SOA in the atmosphere.

In addition, air visibility greatly improved with decline of PM_{2.5} chemical components. We used a local parameter scheme to calculate b_{ext} and demonstrated that it decreased at a rate of $-21.44 \text{ Mm}^{-1} \text{ yr}^{-1}$. The revised IMRPOVE formula will generate higher b_{ext} values than local parameter scheme under high pollution conditions. Thus, more site-specific parameters and local parameter scheme are needed in polluted areas to predict b_{ext} more accurately.

This study highlights that the changes in PM_{2.5} chemical composition can significantly affect key aerosol physico-chemical properties, such as aerosol pH, ALWC, and light extinction coefficient. The variations in aerosol pH and ALWC can, in turn, influence the formation of secondary species. Since the importance of secondary species will become more prominent with continuous air quality improvement, more efforts should focus on them in the future.

Code and data availability. The experimental data in this study are available upon request to the corresponding author by email.

Supplement. The supplement related to this article is available online at <https://doi.org/10.5194/acp-25-13729-2025-supplement>.

Author contributions. XD conceived the project and designed the study. YFH performed the data analysis and wrote the paper. YQZ, DHC, TZ, KY, JQW, QC, and HJ arranged the sample collection and assisted with the data analysis. ZRW and PL analyzed the samples. XD, QFH, XMW, and MB performed the data interpretation and edited the paper. All authors contributed to the development of the final paper.

Competing interests. At least one of the (co-)authors is a member of the editorial board of *Atmospheric Chemistry and Physics*. The peer-review process was guided by an independent editor, and the authors also have no other competing interests to declare.

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