

Research Article

Physicochemical Characteristics of PM_{2.5} Based on Long-term Hourly Data at National Intensive Monitoring Sites in Korea

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ABSTRACT The objective of this study was to analyze the temporal and spatial characteristics of fine particulate matters by using huge hourly datasets of PM_{2.5}, including chemical information monitored at the 6 national intensive monitoring sites (NIMs) from 2013 to 2018 in Korea. Hourly PM_{2.5} raw datasets were obtained from the National Institute of Environmental Research (NIER) in Korea. Monitoring sites included urban, rural/agricultural, industrial, and marine environments. Since the PM_{2.5} concentration steadily decreased nationwide, each species concentration also decreased in general. One of key reasons for decreasing PM_{2.5} might be explained by the implementation of domestic fine dust reduction policies and external influences such as PM_{2.5} concentration reduction in China. It was observed that 45.0% of all datasets for 6 years were classified as good condition. The average sum of 14 elements over all sites in 2018 was calculated to be 501.5 ng/m³, and its mass ratio for PM_{2.5} (21.9 µg/m³) was 2.30%. The inorganic elements were generally higher in industrial/urban areas than in agricultural areas. In addition, the average TC (total carbon) over all 6 sites was 28.3% of PM_{2.5} with the range of 23.6% to 31.4%. The TC in small urban areas was much higher than that in marine areas or even that in large, populated urban area/industrial areas. It seemed that the latter areas were better controlled than the former area in terms of combustion activities of fossil fuels. It is suggested that these results could be play an important role as important basic data to manage ambient air quality and establish effective emission reduction strategies in each region.

KEY WORDS PM_{2.5}, PM₁₀, Chemical composition, Mass ratio, Unidentified fraction

1. INTRODUCTION

Fine particulate matter (FPM) in the accumulation size mode remains and is transported in air much longer than that in other modes (McMurry *et al.*, 2004; EPA, 1999). Thus, it can increase exposure time and directly affect health risks. Further, it is obvious that PM penetrates deep into the alveoli through the upper and lower respiratory tracts of the human body and causes pulmonary diseases such as emphysema or asthma, or cardiovascular diseases (Dockery and Stone, 2007). In addition, it reduces visual range and adversely affects crop growth (KOSAE, 2018). The WHO (World Health Organization) estimated that PM in the air environment

contributed to 3.2 million premature deaths worldwide in 2010, mostly from cardiovascular diseases, and 223,000 deaths from lung cancer (WHO, 2013). Related to outdoor and indoor air pollution, WHO recently evaluated 7 million deaths worldwide in 2016 due to cardiovascular disease, stroke, respiratory disease and cancer (WHO, 2021a), and then strengthened air quality guidelines for $PM_{2.5}$ down to $5 \mu\text{g}/\text{m}^3$ for the annual mean and $10 \mu\text{g}/\text{m}^3$ for the 24-hr mean in Sep. 2021 (WHO, 2021b).

The National Ambient Air Quality Standard (NAAQS) in Korea for PM started in 1983 as total suspended particulate (TSP) with a diameter of $500 \mu\text{m}$ or less. TSP was deleted from the NAAQS, and only PM_{10} has been regulated since 2001 (Kim, 2013). In 2015, the NAAQS for $PM_{2.5}$ was set to an annual mean of $25 \mu\text{g}/\text{m}^3$ and a 24-hr mean of $50 \mu\text{g}/\text{m}^3$. However, since the standard was too loose compared to WHO or other foreign standards, the MOE (Ministry of Environment) strengthened the $PM_{2.5}$ standard a few years later, and the annual and 24-hr standards have been changed to $15 \mu\text{g}/\text{m}^3$ and to $35 \mu\text{g}/\text{m}^3$, respectively, since March 2018 (MOE, 2018). According to a report (MOE, 2020a), annual average level for $PM_{2.5}$ slightly improved to $26 \mu\text{g}/\text{m}^3$ in 2016, $25 \mu\text{g}/\text{m}^3$ in 2017, and $23 \mu\text{g}/\text{m}^3$ in 2018, but the levels still far exceeded the latest standard of $15 \mu\text{g}/\text{m}^3$. Due to significant social problems related to $PM_{2.5}$ in the last decade in Korea, the “Special Act on PM Reduction and Management” was enacted on Feb 2019 (MOE, 2019a). By law, the MOE had to prepare a comprehensive PM management plan by November 2019. The goal of the plan was to reduce the average level of $PM_{2.5}$ nationwide from $26 \mu\text{g}/\text{m}^3$ in 2016 to $16 \mu\text{g}/\text{m}^3$ in 2024 during the period of 2020 to 2024 (MOE, 2019b).

PM is not a simple compound, but a mixture of particles with various sizes containing various chemical species. Usually, PM is composed of sulfate, nitrate, ammonium, and other ionic species, carbon species (such as organic, inorganic, and elemental carbons) and other inorganic elements. Properly providing supplementary meteorological parameters, the chemical species in PM (PM_{10} and $PM_{2.5}$) provide key information on where the PM came from, what the PM sources were, how the PM formed, and even when the PM formed. For example, PMs composed of many elements such as Si, Al, Fe, K, Ca are mainly emitted from a soil-related source, sulfate is from combusting sulfur-containing fossil fuels, and V and Ni are from oil combustion sources (Hopke, 1985).

Recently, there have been many studies to characterize chemical species of $PM_{2.5}$ nationwide in Korea such as a study of chemical characteristics of $PM_{2.5}$ in Seoul in 2019 (Um *et al.*, 2020), a study of the origins and distributions of atmospheric ammonia in Jeonju during 2019–2020 (Park *et al.*, 2020), a study on the characteristics of $PM_{2.5}$ chemical compositions and high-concentration episodes from 2013 to 2016 in Jeju (Kim *et al.*, 2020), a study on $PM_{2.5}$ and its chemical compositions from 2017 to 2018 in Jeonju (Jo *et al.*, 2018), estimation of emission source contribution of OC and EC in the spring of 2016 in Seoul (Ham *et al.*, 2017), and a study on the distribution of heavy metals during 2013 and 2014 in Gyeonggi-do (Kim *et al.*, 2014).

The purpose of this study is to analyze the temporal and spatial characteristics of fine particulate matters by using huge hourly datasets of $PM_{2.5}$, including chemical information monitored at the 6 national intensive monitoring sites (NIMSs) in Korea during the period of 2013 to 2018. These research results can play an important role as basic data for air quality management and establishment of effective emission reduction strategies in each region.

2. EXPERIMENTAL METHODS

2.1 Data Collection and Analytical Methods

The national intensive monitoring sites (NIMSs) have been operated by the Ministry of Environment (MOE) since 2013 and were developed (i) to describe the status of the ambient air environment nationwide including urban/agricultural/industrial/marine areas, (ii) to measure various ionic species and trace elements contained in long-range transported PM, and finally (iii) to identify the reasons for high PM episodes in Korea. There are a total of six sites such as the Joongbu Site (NIMS-JB), the Seoul Metropolitan Site (NIMS-SM), the Honam Site (NIMS-HN), the Baengnyeongdo Site (NIMS-BN), the Youngnam Site (NIMS-YN), and the Jeju Site (NIMS-JJ) (MOE, 2019c). Recently, the MOE has made the hourly raw data containing chemical information of $PM_{2.5}$ available. Fig. 1 shows the locations of these six NIMSs in Korea, and Table 1 provides information on location, site category, and monitoring purpose for each NIMS (NIER, 2019). In summary, NIMS-SM and NIMS-JB were established to represent air in large and medium urban areas, respectively, NIMS-HN is used to monitor

rural/agricultural areas, NIMS-YN to monitor industrial air, and NIMS-BN and NIMS-JJ to monitor background marine air.

The MOE provided raw sample datasets monitored during 6 years from January 1, 2013 to December 31, 2018 for all 6 NIMSs. The hourly datasets for PM₁₀ and PM_{2.5} as well as chemical information on PM_{2.5} were analyzed in this study. PM₁₀ and PM_{2.5} were separately monitored at each NIMS via size selection by an inertial

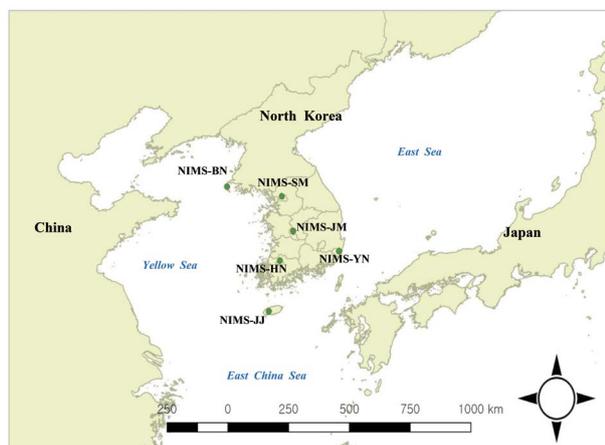


Fig. 1. Location of six national intensive monitoring sites in Korea.

impactor type inlet and a cyclone type inlet, respectively. After collecting the PM on each filter, mass concentration was determined by a beta-ray absorption method, which detects beta-rays before and then determines PM concentration by beta-ray attenuation after sampling with a high-sensitivity detector (NIER, 2019). Ionic species in PM_{2.5} were analyzed by ion chromatography after PM collection. Carbon species were analyzed by a thermal/optical transmittance method and a non-dispersive infrared method. Finally, non-destructive x-ray fluorescence spectroscopy (XRF) was employed for trace elements. Further detailed analytical and sampling methods in this study can be found in the literature (NIER, 2019).

2.2 Data Pretreatment Steps for Characterization of PM and Chemical Species in PM_{2.5}

Statistical analysis was performed after proper pretreatments for sample datasets monitored at hourly intervals from 2013 to 2018 over 6 NIMSs. Initially when either a PM₁₀ or PM_{2.5} mass datum was missing, it was removed from each sample dataset at each NIMS. A datum was also removed when the PM_{2.5} mass was higher than the PM₁₀ mass, that is, when F/C ratio > 1 (i.e., when the mass ratio of fine particles/coarse particles = PM_{2.5}/

Table 1. NIMS information in Korea (NIER, 2019).

Site name (Code)	Site category	Year opened	Location	Longitude	Latitude	Monitoring purposes
Joongbu (NIMS-JB)	Medium urban	2010	Daejeon City	127.41	37.61	- To monitor urban air pollutants - To identify the impact of local emission and long-range transport on local air quality
Seoul Metropolitan (NIMS-SM)	Large urban	2008	Seoul City	126.94	37.61	- To monitor urban air pollutants - To identify the impact of local emission and long-range transport on local air quality
Honam (NIMS-HN)	Small urban/ Rural/ Agricultural	2009	Gwangju City	126.85	35.23	- To monitor urban and rural/agricultural air pollutants
Baengnyeong (NIMS-BN)	Marine in small island	2007	Baengnyeong Island	124.63	37.97	- To monitor background air quality on the Western coast of Korea - To identifying the impact of long-range transported air pollutants from the outside
Youngnam (NIMS-YN)	Industrial	2013	Ulsan City	129.32	35.58	- To monitor air pollutants from the Ulsan industrial complex
Jeju (NIMS-JJ)	Marine in large island	2012	Jeju Island	126.39	33.35	- To monitor background air quality on the Southern coast of Korea - To identify the impact of long-range air pollutants transported from outside

Table 2. Numbers of sample datasets deleted and used at each NIMS after data pretreatment.

Pretreatment step		NIMS-JB	NIMS-SM	NIMS-HN	NIMS-BN	NIMS-YN	NIMS-JJ	ΣNIMS
Study-1: The 1 st Pretreatment step for PM ₁₀ and PM _{2.5} measured in 2013 to 2018	Raw datasets	52,584	51,048	52,584	52,584	43,824	52,584	305,208
	Datasets deleted	3,724	4,861	4,478	5,304	2,698	6,462	27,527
	Datasets used	48,860	46,187	48,106	47,280	41,126	46,122	277,681
	Ratio (used/total)	92.9%	90.5%	91.5%	89.9%	93.8%	87.7%	91.0%
Study-2: The 2 nd Pretreatment step for 8 ions and 2 carbons measured in 2013 to 2018	Raw datasets	52,584	51,048	52,584	52,584	43,824	52,584	305,208
	Datasets deleted	23,479	29,127	16,892	22,417	19,205	27,757	138,877
	Datasets used	29,105	21,921	35,692	30,167	24,619	24,827	166,331
	Ratio (used/total)	55.3%	42.9%	67.9%	57.4%	56.2%	47.2%	54.5%
Study-3: The 3 rd Pretreatment step for 8 ions, 2 carbons and 14 elements measured in 2018	Raw datasets	8,760	8,760	8,760	8,760	8,760	8,760	52,560
	Datasets deleted	2,988	4,220	2,833	4,010	5,791	4,808	24,650
	Datasets used	5,772	4,540	5,927	4,750	2,969	3,952	27,910
	Ratio (used/total)	65.9%	51.8%	67.7%	54.2%	33.9%	45.1%	53.1%

PM₁₀ ratio is over 1). After performing the 1st pretreatment step for Study-1, 277,681 (91.0%) out of 305,208 raw sample datasets were left as shown in Table 2. By using the remaining samples, PM₁₀ and PM_{2.5} mass behaviors, F/C ratios, and relationships with meteorological parameters were investigated for each site.

In Study-2 for chemical characterization, all raw datasets consisting of PM_{2.5} mass, 8 ionic species (SO₄²⁻, NO₃⁻, NH₄⁺, Cl⁻, Na⁺, K⁺, Mg²⁺, and Ca²⁺), and 2 carbon components (OC, EC) were examined during the 2nd pretreatment step. When any one of those 10 chemical species was missing or if the sum of them was greater than PM_{2.5} mass, the hourly dataset for that time was deleted from the analysis. The number of raw samples was 305,208, but only 166,331 samples (54.5%) remained after the pretreatment. In Study-3, the mass ratio for each inorganic element was analyzed based on samples measured only in 2018. Among the initial raw data of 17 inorganic elements (i.e., Si, S, K, Ca, Ti, V, Cr, Mn, Fe, Ni, Cu, Zn, As, Se, Br, Ba, and Pb), three elements (S, K, and Ca) were excluded since those overlapped with ion species mentioned above. For the 3rd data pretreatment step, all the corresponding raw samples were deleted during this pretreatment step if 1) any of the 24 species including 14 elements was missing in the same sam-

ple, 2) PM_{2.5} measurement was missing or its mass concentration was zero, or 3) if the F/C ratio exceeded 1. After the step, the number of initial samples was 52,560, but only 27,910 samples (53.1%) were left for the next analysis.

3. RESULTS AND DISCUSSION

3.1 Physical Characteristics of PM (PM₁₀ and PM_{2.5}) for Study-1

For 6 years from 2013 to 2018, the average annual range of PM₁₀ concentration at each NIMS was 28.9 to 50.9 µg/m³, and PM_{2.5} was 16.1 to 31.9 µg/m³. Table 3 summarizes 6-year average statistics for PM, F/C ratio, and meteorological parameters at each site. For the PM levels, it was observed that NIMS-JB > NIMS-SM > NIMS-HN > NIMS-BN > NIMS-YN > NIMS-JJ. The NIMS-JB located on a medium sized urban city in the middle of Korea showed the highest PM level. However, the NIMS-JJ located on the biggest southern island showed the lowest PM level. As shown in Table 3, PM_{2.5} levels over the country exceeded the 2021 WHO guidelines as well as the 2018 Korean NAAQS.

In our study, the average range of the F/C ratio was

Table 3. A summary of statistics for 6-year averages of PM, F/C ratio, and meteorological parameters (Study-1).

Site ID	PM ₁₀ ($\mu\text{g}/\text{m}^3$)	PM _{2.5} ($\mu\text{g}/\text{m}^3$)			F/C ratio	Temp ($^{\circ}\text{C}$)	Precip. (mm/yr)	RH (%)	WS (m/s)
		Average	StDev	Max					
NIMS-JB (medium urban air)	50.9	31.9	21.8	204.0	0.62	13.6	1,159.2	70.4	1.35
NIMS-SM (large urban air)	48.6	30.0	22.4	205.0	0.64	13.1	1,086.0	59.5	2.11
NIMS-HN (rural/agricultural air)	44.4	27.1	19.2	165.0	0.62	14.6	1,238.4	68.6	1.51
NIMS-BN (western marine air)	42.5	23.0	20.8	251.0	0.55	11.5	626.4	71.2	4.10
NIMS-YN (industrial air)	38.2	21.4	15.6	143.0	0.56	14.7	1,244.4	65.9	2.17
NIMS-JJ (southern marine air)	28.9	16.1	14.7	183.0	0.56	15.8	1,168.8	77.4	3.08

0.55 to 0.64 over 6 NIMSs, as shown in Table 3. We observed that the F/C ratio tended to be higher as the average concentrations of PM₁₀ and PM_{2.5} increased together. This means that the increase rate of PM_{2.5} is generally faster than that of PM₁₀. Further, the ratio was higher in urban and agricultural areas than in marine and industrial areas. It is obvious that there are enormous PAL sources (point, area, line source) emitting primary PM_{2.5} as well as man-made gaseous pollutants forming secondary PM_{2.5} in densely populated urban areas. However, there are few distinct PAL sources in sparsely populated background marine areas. Though many point sources exist in industrial areas (like near NIMS-YN), it seems that the sources in that area are well controlled by strict emission regulations. On the other hand, even though there are few point and line sources in agricultural areas like near NIMS-HN, the area sources emitting gaseous pollutants like NH₃ might be a major contributor to the formation of secondary PM_{2.5}, and there are many scattered area sources emitting various gaseous and particulate matter from illegal burning of agricultural wastes and solid refused fuel (SRF) in that area. According to the CAPSS (Clean Air Policy Support System) data of the Ministry of Environment as of 2018, NH₃ emissions from the agricultural sector were the highest in Honam region with 67,836 tons, followed by the Youngnam, Joongbu, and the metropolitan area at 59,930 ton, 59,877 ton, and 42,677 tons, respectively (NAEIRC, 2019).

A correlation result among annual averages of PM₁₀,

PM_{2.5}, and F/C ratios showed that PM₁₀ and PM_{2.5} were highly correlated with $r = 0.941$ and $p\text{-value} = 0.000$. However, PM_{2.5} and F/C ratio were moderately correlated with $r = 0.737$ and $p\text{-value} = 0.000$. Furthermore, PM₁₀ and F/C ratios were poorly correlated with $r = 0.513$ and $p\text{-value} = 0.002$. In addition, PM₁₀ and PM_{2.5} were higher at NIMS-JB and NIMS-SM, both located in urban areas, where WS was quite low at 1.35 m/s and 2.11 m/s, respectively, than at NIMS-JJ (in a marine area), where PM₁₀ and PM_{2.5} were the lowest with an average WS of 3.08 m/s. Thus, it seems to be that WS inversely impacted PM₁₀ and PM_{2.5} by dispersed bulk motion. However, even though NIMS-BN located in far western marine areas had the highest wind speed of 4.1 m/s, elevated levels of PM₁₀ and PM_{2.5} were observed compared to NIMS-YN in land industrial area and NIMS-JJ in a southern marine area. It might be directly affected by PM flowing from outside since NIMS-BN is 180 km away from mainland China and only 14 km away from North Korea. According to Im *et al.* (2021), high PM₁₀ and PM_{2.5} during the period of 2015 to 2020 were affected by incoming PM from outside because there are no distinct emitting sources inside the island.

As another correlation result among annual averages of PM₁₀, PM_{2.5}, and WS, when all average data over all 6 sites were used, both PM₁₀ and PM_{2.5} were weakly but negatively correlated with WS with $r = -0.436$, $p\text{-value} = 0.009$ and $r = -0.518$, $p\text{-value} = 0.001$, respectively, at 95% significance level, as shown in Fig. 2(a) and (b).

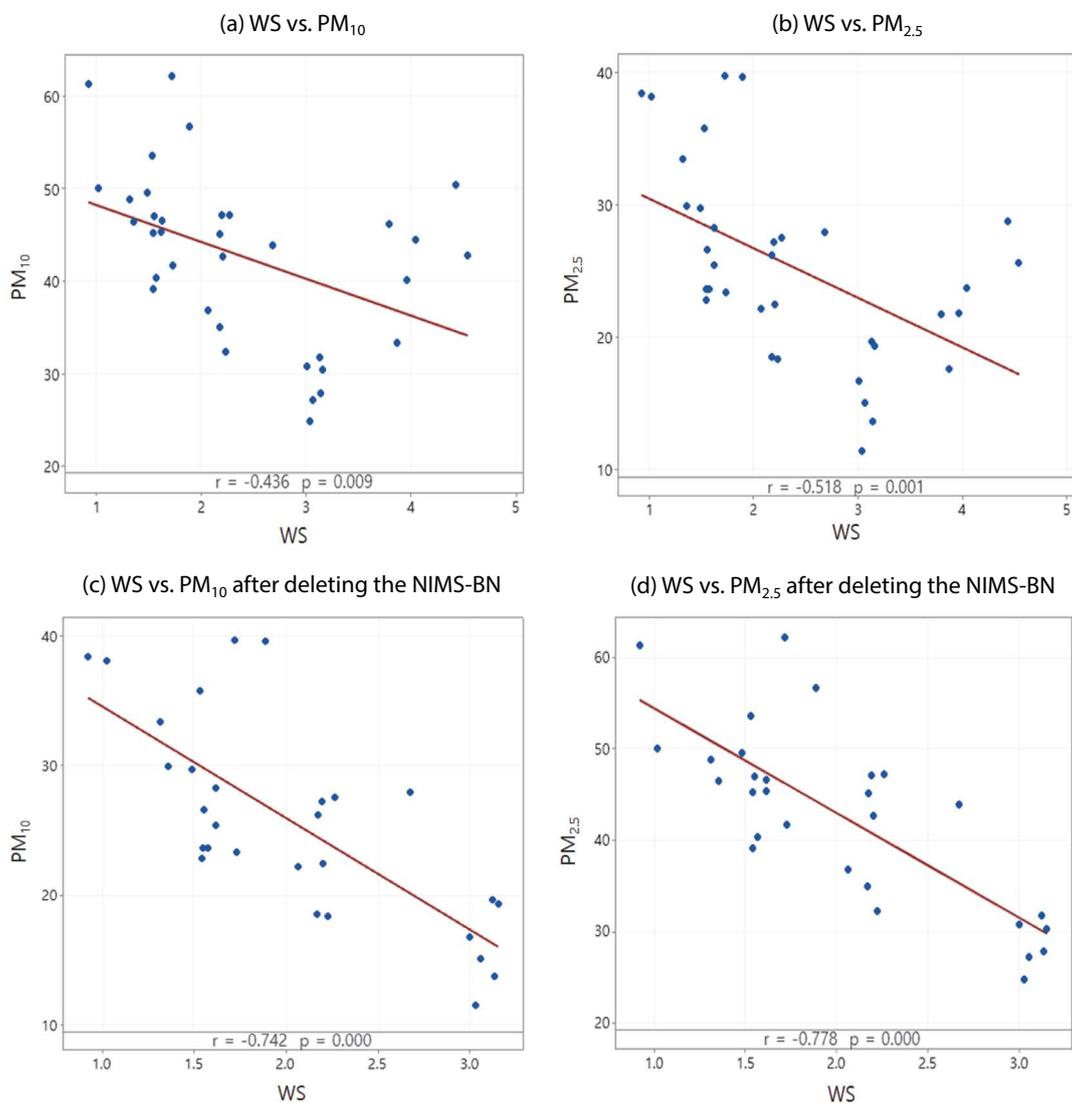


Fig. 2. Scattered correlation plots for WS vs. PM_{10} and WS vs. $PM_{2.5}$: (a) WS vs. PM_{10} , (b) WS vs. $PM_{2.5}$, (c) WS vs. PM_{10} after deleting samples of NIMS-BN, and (d) WS vs. $PM_{2.5}$ after deleting samples of NIMS-BN.

However, since NIMS-BN had a big standard deviation and the fastest WS when calculating annual averages of PM, we reanalyzed the relationships between PM and WS after excluding NIMS-BN. The result showed that the correlation coefficient was negative but stronger than before, as shown in Fig. 2(c) and (d). PM_{10} and $PM_{2.5}$ were clearly negatively correlated with WS by $r = -0.742$, $p\text{-value} = 0.000$ and $r = -0.778$ and $p\text{-value} = 0.000$, respectively. The point was that less scattered plots were obtained when NIMS-BN was excluded in the analysis. The air quality of NIMS-BN was likely influenced by wind direction rather than wind speed because the site is

located on a small island is surrounded by China, South and North Korea and it has few internal emission sources, as mentioned above.

Fig. 3 shows annual average concentrations of PM_{10} and $PM_{2.5}$ at each NIMS, which show an overall decreasing trend. It must be noted that the annual average PM was calculated with sample datasets after performing the data pretreatment step, as shown in Table 2. Thus, 277,681 out of 305,208 datasets were utilized in this analysis. As shown in Fig. 3, the annual PM_{10} and $PM_{2.5}$ at NIMS-JB gradually decreased after peaking in 2014 at $61.3 \mu\text{g}/\text{m}^3$ and $38.4 \mu\text{g}/\text{m}^3$, respectively. They were 45.3

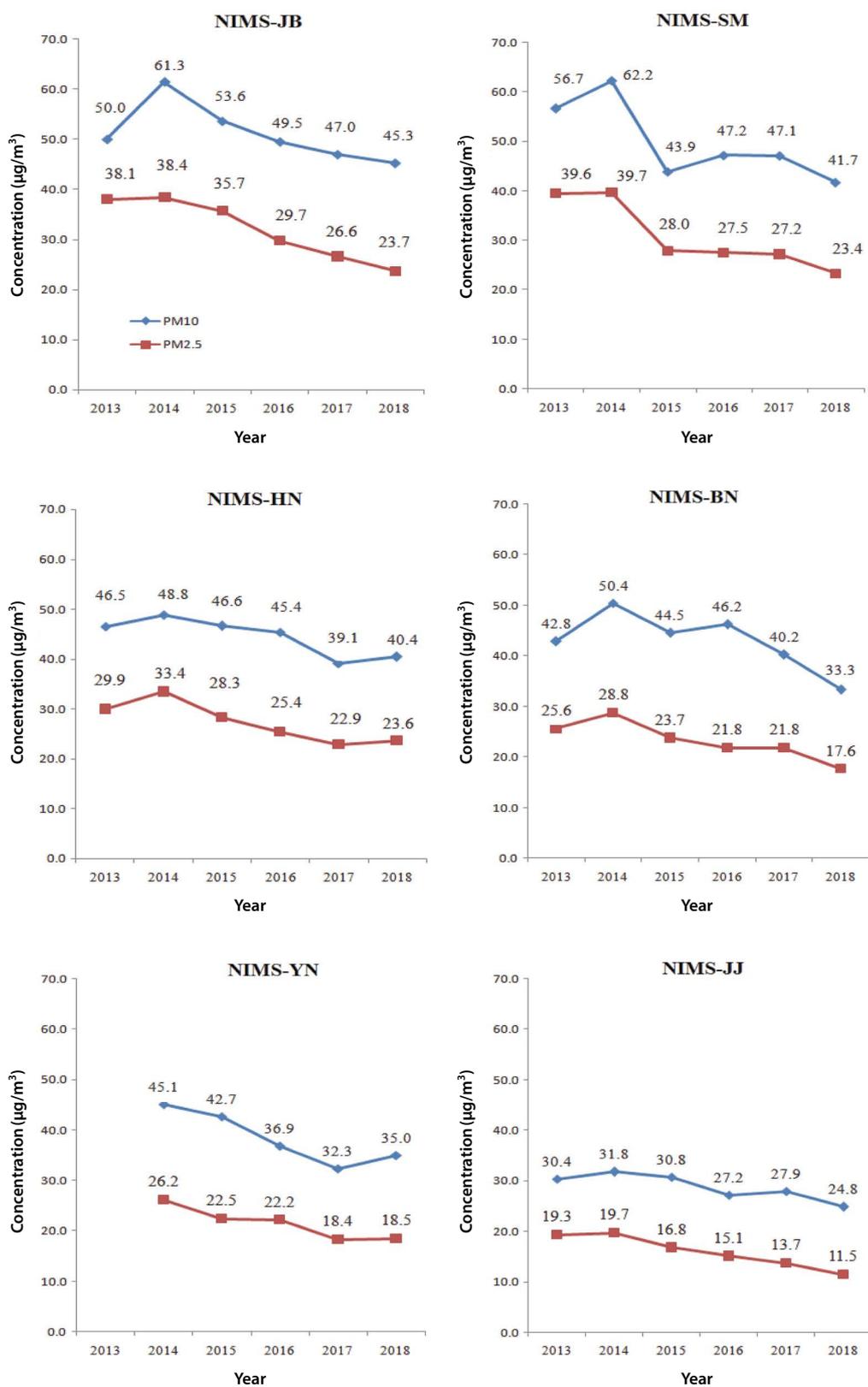


Fig. 3. Trends of annual average PM₁₀ & PM_{2.5} concentrations at the 6 NIMSs during Study-1.

$\mu\text{g}/\text{m}^3$ and $23.7 \mu\text{g}/\text{m}^3$ in 2018, respectively, and decreased by 9.5% and 37.8%, respectively, compared to 2013. For a case at NIMS-SM in Seoul, PM_{10} and $\text{PM}_{2.5}$ in 2018 decreased by 26.5% and 41.0%, respectively, compared to 2013. The PM_{10} and $\text{PM}_{2.5}$ of NIMS-JJ on Jeju Island decreased by 18.2% and 40.5%, respectively, compared to 2013. The NIMS-JJ was only one site that could meet the new $\text{PM}_{2.5}$ standard of $15 \mu\text{g}/\text{m}^3$ in the years of 2017 and 2018. We suggest that a reason for this decreasing trend of PM is the change in socio-environmental points of view in Korea. It seems that the social concern and activity for $\text{PM}_{2.5}$ soared since 2013, even though the MOE started to measure $\text{PM}_{2.5}$ nationwide in the beginning of 2015. In 2013, a pilot forecast of $\text{PM}_{2.5}$ was implemented after the International Agency for Research on Cancer (IARC) classified air pollution and $\text{PM}_{2.5}$ as group 1 carcinogens in 2013. Since then, social interests in PM, especially $\text{PM}_{2.5}$, have increased considerably. The government had to implement and enforce various laws and policies every year from then on, such as the Special Plan on PM Management in 2016, the Comprehensive Plan on PM Management in 2017, the Special Act on PM Reduction and Management in 2019, and so forth (MOE, 2019b) to deal with public complaints. It is estimated that the continuous decrease in $\text{PM}_{2.5}$ concentration has improved further due to external influences such as fine dust reduction in China as well as the implementation of domestic fine dust reduc-

tion policies. The $\text{PM}_{2.5}$ concentration in Beijing, China decreased by about 50.1% from $101.6 \mu\text{g}/\text{m}^3$ in 2013 to $50.7 \mu\text{g}/\text{m}^3$ in 2018 (Statista, 2022).

To comprehensively understand the air quality status, we initially analyzed all 277,681 hourly $\text{PM}_{2.5}$ samples over 6 NIMSS and classified the samples into 5 conditions according to the forecast criteria by the guideline of the MOE (MOE, 2018): good condition ($\text{PM}_{2.5} \leq 15 \mu\text{g}/\text{m}^3$), moderate ($15-35 \mu\text{g}/\text{m}^3$), bad ($35-75 \mu\text{g}/\text{m}^3$), very bad ($75-150 \mu\text{g}/\text{m}^3$), and the worst warning class ($\text{PM}_{2.5} > 150 \mu\text{g}/\text{m}^3$). As a result, the air quality condition for 6 years is shown in Fig. 4. The results were as follows: 45.0% of the hourly $\text{PM}_{2.5}$ samples were sorted into good condition, 38.0% in normal, 15.2% in bad, 1.8% in very bad, and 0.1% in the worst warning conditions. The figure shows the trend of the conditions for each NIMS and plots (a) through (f) contain annual trends for each NIMS. The sums of bad and worse conditions (that is, highly polluted conditions) were 34.6%, 30.0%, and 25.3% observed at NIMS-JB, NIMS-SM, and NIMS-HN, respectively, while 17.4% and 16.4% were observed at NIMS-BY and NIMS-YN, respectively, and lastly NIMS-JJ was only 8.8%. It is obvious that the good/moderate conditions are increasing in all sites, and the bad/very bad/worst conditions are continuously decreasing. Specifically, in the case of NIMS-SM in Seoul, the percentage of good/normal conditions in 2018 increased by 82.4% compared to 2013. The improved

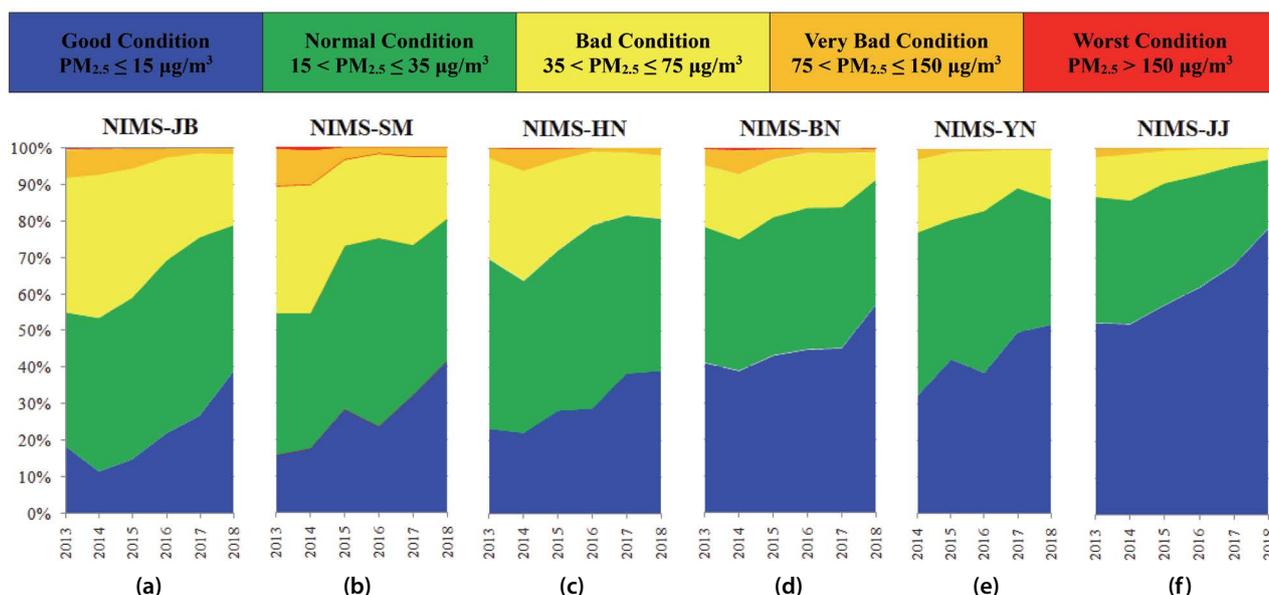


Fig. 4. Annual trends of $\text{PM}_{2.5}$ air quality conditions.

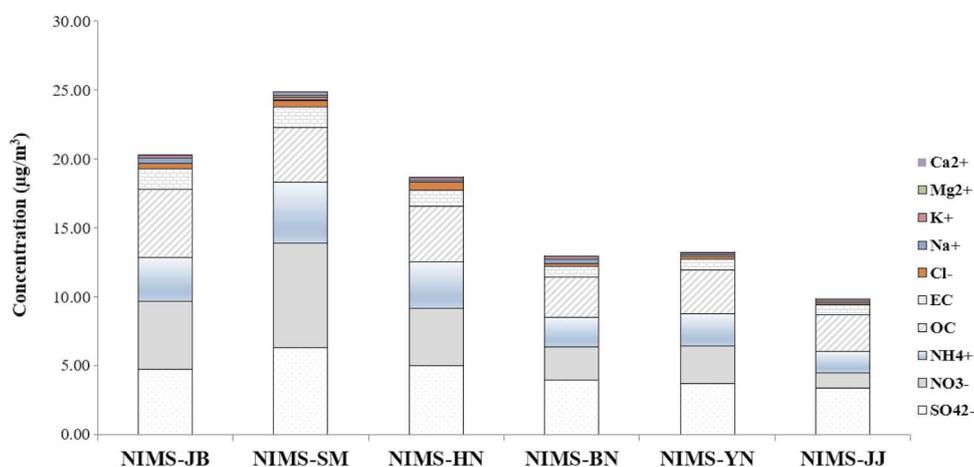


Fig. 5. Stacked histograms showing 6-yr average concentrations for 10 chemical species in PM_{2.5}.

percentages in 2018 compared to 2013 were 37.2% in NIMS-JB, 30.3% in NIMS-YN, 22.8% in NIMS-BN, and 14.4% in NIMS-HN, respectively. The value for NIMS-JJ on Jeju Island, was only 1.4% with no significant difference since relatively good air quality conditions were observed in 2013. Fortunately, the data seldom exceeded 150 µg/m³ for PM_{2.5} except at NIMS-SM in 2017 and NIMS-BN in 2018.

3.2 Chemical Characteristics of 8 Ionic Species and EC/OC in PM_{2.5} for Study-2

PM_{2.5} is emitted directly from various sources as primary aerosols such as PAL sources, which are mostly emitted by combustion of fossil fuels (Kondratyev *et al.*, 2006). PM_{2.5} is also generated as a secondary aerosol in the forms of sulfate (SO₄²⁻), nitrate (NO₃⁻), ammonium (NH₄⁺) by chemical reactions with various precursor gases like sulfur oxides (SO_x), nitrogen oxides (NO_x), and ammonia (NH₃) in the troposphere (EEA, 2020). In the presence of moisture (H₂O) in the atmosphere, the SO₄²⁻ and NO₃⁻ react with NH₃ to generate particulate ammonium sulfate ((NH₄)₂SO₄) and ammonium nitrate (NH₄NO₃). In addition, secondary PM_{2.5} is also generated by a photochemical reaction between ozone (O₃) and various volatile organic carbons (VOC) in the form of secondary organic aerosols (SOAs) in the atmosphere (McMurry *et al.*, 2004).

Among the chemical species measured hourly at 6 NIMSS, the 6-yr average concentrations (5-yr average in the case of NIMS-YN) for each species in PM_{2.5} were calculated and then plotted in Fig. 5. It must be noted

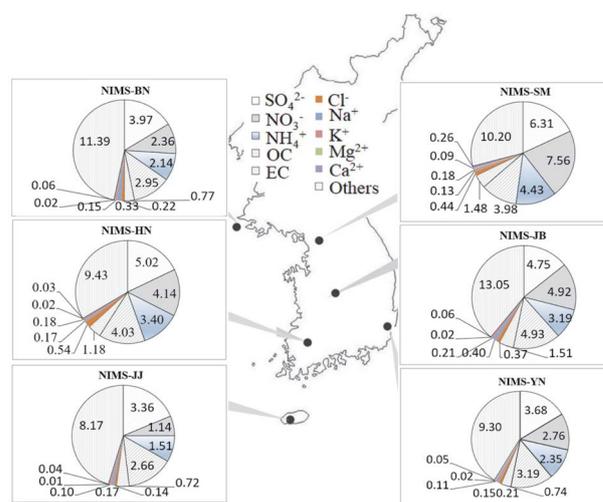


Fig. 6. A site map with pie charts showing 6-yr average compositions for 10 chemical species in PM_{2.5}.

that the average concentration was calculated based on 166,331 samples over 6 NIMSS after performing data pretreatment steps. The figure shows histograms of stacked average concentrations for 10 chemical species including 8 ionic species (SO₄²⁻, NO₃⁻, NH₄⁺, Cl⁻, Na⁺, K⁺, Mg²⁺, and Ca²⁺), and 2 carbon components (OC, EC). Again, each concentration for each species was a mean calculated based on the hourly samples during the period of 2013 to 2018 except NIMS-YN during 2014 to 2018. In addition, Fig. 6 is a location map together with chemical compositions in terms of average ratios of each species to PM_{2.5}. The 'others' in the figure stands for the

average percentage of all other unmeasured species except the 10 measured species in PM_{2.5}.

Table S1 shows the overall summary of Study-2 containing annual average concentrations and mass ratios for five major chemical species such as SO₄²⁻, NO₃⁻, NH₄⁺, OC, EC in PM_{2.5} measured at each NIMS after proper data pretreatment. When employing all 166,331 sample datasets together to obtain basic statistics during the study period, the average species concentration and its mass ratio to PM_{2.5} (26.85 µg/m³) were calculated: SO₄²⁻ (4.48 µg/m³, 16.7%), NO₃⁻ (3.79 µg/m³, 14.1%), NH₄⁺ (2.84 µg/m³, 10.6%), OC (3.65 µg/m³, 13.60%), EC (1.07 µg/m³, 4.0%), Cl⁻ (320 ng/m³, 1.2%), Na⁺ (226 ng/m³, 0.8%), K⁺ (155 ng/m³, 0.6%), Mg²⁺ (27 ng/m³, 0.1%), Ca²⁺ (75 ng/m³, 0.3%), and the rest (10.22 µg/m³, 38.1%). Describing the 6-yr average at each site for each species, average SO₄²⁻ concentrations ranged from 3.36 to 6.31 µg/m³ in the order of NIMS-SM (6.31 µg/m³) > NIMS-HN (5.02 µg/m³) > NIMS-JB (4.75 µg/m³) > NIMS-BN (3.97 µg/m³) > NIMS-YN (3.68 µg/m³) > NIMS-JJ (3.36 µg/m³).

The average mass ratio of SO₄²⁻ was the highest among all 10 species measured. The SO₄²⁻ ratios at NIMS-SM and NIMS-HN were higher than the other NIMSs. It is known that SO₂ as a precursor of SO₄²⁻ is emitted from power plants, industrial facilities, and various fugitive area sources using fossil fuels such as coal, oil, and bio-SRF (biomass-solid refuse fuel) containing sulfur. Even though the forming rate in winter is slower than in other seasons (Hodan and Barnard, 2004), a huge amount of SO₄²⁻ was emitted from local areas in addition to long-range transport from the outside regional area due to an increase in burning activity in winter (Park *et al.*, 2019). The Korean MOE reported that SO₂ is mainly emitted from the power generation sector in the energy industry and other sectors in petroleum product and manufacturing industries (MOE, 2020a). The European Environment Agency (EEA) estimated that SO₂ is emitted from energy supply (47%), manufacturing and industrial sectors (33%) (EEA, 2020).

As shown in Fig. 5, the average concentrations of NO₃⁻ in 2018 ranged from 1.14 to 7.56 µg/m³ over NIMSs, in the order of NIMS-SM (7.56 µg/m³) > NIMS-JB (4.92 µg/m³) > NIMS-HN (4.14 µg/m³) > NIMS-YN (2.76 µg/m³) > NIMS-BN (2.36 µg/m³) > NIMS-JJ (1.14 µg/m³). The average mass ratio of NO₃⁻ in PM_{2.5} was 14.2%, showing the second highest ratio among 10 species. Particularly, NIMS-SM in Seoul was observed to

have a 21.5% mass ratio, which was remarkably high compared to the other NIMSs. It is obvious that gaseous NO and NO₂ as a precursor of particulate NO₃⁻ are widely emitted from mobile sources in metropolitan areas. In the USA, SO₄²⁻ was higher in the eastern part of the country than in the west, whereas NO₃⁻ was a specific major pollutant in western urban areas (Hodan and Barnard, 2004). The EEA showed that NO_x was emitted mainly from mobile sources (39%), energy supply, manufacturing and industrial sectors, and agriculture (15% each) (EEA, 2020). According to the CAPSS in Korea (MOE, 2020b), mobile sources including passenger cars were the largest emitters, followed by freight cars > ships > RVs > construction equipment. Point sources such as power plants and chemical manufacturing processes were the next highest emitter of NO_x.

The average NH₄⁺ in Fig. 6 ranged from 1.51 to 4.43 µg/m³ in the order of NIMS-SM (4.43 µg/m³) > NIMS-HN (3.40 µg/m³) > NIMS-JB (3.19 µg/m³) > NIMS-YN (2.35 µg/m³) > NIMS-BN (2.14 µg/m³) > NIMS-JJ (1.51 µg/m³). The average mass ratio over all NIMSs was 10.5% of PM_{2.5}, and NIMS-SM and NIMS-HN were respectively 12.6% and 12.1% higher than the other NIMSs. In the USA, gaseous NH₃ as the precursor of NH₄⁺ is mostly emitted from agricultural activities, urban/commercial composting processes, mobile sources, and particular chemical industries (EPA, 2021). The EEA estimated that most NH₃ (93%) was emitted from agricultural activities (EEA, 2020). In Korea, even though emission inventory for NH₃ is not yet complete, it seemed to be mostly emitted from manure processing and fertilizer use in the agricultural sector as well as from the petroleum product industry sector (MOE, 2020b). As a result, the average organic carbon (OC) in Fig. 6 was in the range of 4.93–2.66 µg/m³, in the order of NIMS-JB (4.93 µg/m³) > NIMS-HN (4.03 µg/m³) > NIMS-SM (3.98 µg/m³) > NIMS-YN (3.19 µg/m³) > NIMS-BN (2.95 µg/m³) > NIMS-JJ (2.66 µg/m³). It occupied the third highest mass ratio, representing 13.5% of PM_{2.5}. On the other hand, elemental carbon (EC) was 1.51–0.72 µg/m³ in the order of NIMS-JB (1.51 µg/m³) > NIMS-SM (1.48 µg/m³) > NIMS-HN (1.18 µg/m³) > NIMS-BN (0.77 µg/m³) > NIMS-JJ (0.74 µg/m³) > NIMS-YN (0.72 µg/m³). The OC accounted for 3.9% in PM_{2.5}, showing the lowest mass ratio among SO₄²⁻, NO₃⁻, NH₄⁺, OC, and EC. In general, OC is directly emitted by combustion of fossil fuels (as primary source) and is also formed during the condensation of low vapor pressure

products generated from photochemical reactions of VOC as a secondary source. In urban areas, both OC and EC were emitted directly from fossil fuels such as diesel vehicles and bio-incineration (Seinfeld and Pandis, 2012; Hodan and Barnard, 2004). Elevated OC was observed during hazy days in 2001 from biomass burning outside Seoul (Kang *et al.*, 2004). In the case of Europe, more than half the emissions were estimated to be from the manufacturing and industrial sectors (52%), the agricultural sector (20%) and the residential/commercial sector (14%) (EEA, 2020). Although there are no source inventories for EC and OC in PM_{2.5} for recent years in Korea, an inventory of VOCs indicate it is indirectly generated from organic solvent-using sector (MOE, 2020b).

The OC/EC ratios were calculated for all samples over all sites, and the ratios were in the range of 2.86 for the large urban NIMS-SM to 4.14 at the industrial NIMS-YN with an average of 3.56 in Korea. Tan *et al.* (2009) reported that OC/EC ratios ranged from 2.8 to 6.2 with an average of 4.7 during normal and hazy days in Guangzhou, China. Part *et al.* (2001) measured ratios at the Sihwa industrial area in 1998–1999 in Korea. The ratios ranged from 4.4 to 12.0 and were higher than those for other urban and rural environments. They concluded that the ratio from primary emissions was influenced by temporal fluctuations of factory activities in the industrial complex and meteorological conditions. On the other hand, average OC/EC ratio was 6.6, a ratio similar to that of biomass burning emissions in Agra, India (Pachauri *et al.*, 2013).

Again, as can be seen in Fig. 6, the average concentration for other minor species such as Cl⁻, Na⁺, K⁺, Mg²⁺, and Ca²⁺ was in the range of 0.01 to 0.54 µg/m³, accounting for only 0.1 to 1.9% of the mass ratio in PM_{2.5}. The order of their mass ratios was Cl⁻ > Na⁺ > K⁺ > Ca²⁺ > Mg²⁺. Their annual average trends for each NIMS is shown in Fig. 7. Overall, PM_{2.5} concentrations (i.e., the dotted line in the figure), tend to decrease with small fluctuations. SO₄²⁻ at NIMS-YN and NIMS-JJ, NO₃⁻ at NIMS-SM and NIMS-HN, NH₄⁺ at NIMS-YN and NIMS-JJ, and OC at NIMS-SM increased, but all the other species showed stagnant or decreasing trends at all sites. Especially, NO₃⁻ at NIMS-SM in Seoul increased significantly in 2018. However, SO₄²⁻ in 2018 decreased at all 5 sites except at NIMS-YN, where it increased by 15.4% in 2018 as compared to 2013. NO₃⁻ in 2018 was increased by 8.4% to 33.4% at all 5 sites except NIMS-SM as compared to 2013. Also, OC in 2018 was increa-

sed by 5.0% and 0.9% at NIMS-SM and NIMS-HN, respectively. With the ratio of NO₃⁻ to SO₄²⁻ (N/S ratio), the contribution of mobile and stationary pollutants in the air can be assessed qualitatively. In this study, the average N/S ratio was 0.35 to 1.29, which was high in the Seoul metropolitan and Joongbu area, and low in Baengnyeong-do and Jeju-do. The N/S ratio in the Seoul metropolitan area continued to increase except in 2015, and it showed a maximum value of 2.0 in 2018. This means that the Seoul metropolitan area and the Joongbu area have a large effect of mobile emission sources, and it can be confirmed that the fixed pollution source has a large effect on Baengnyeongdo and Jeju Island, which has a relatively small population. The impact of major pollutants by region requires a more in-depth analysis through an receptor model analysis in the future.

Fig. 8 shows stacked plots containing annual trends for 10 chemical compositions in PM_{2.5} at each NIMS. After calculating mass ratios for each species to PM_{2.5}, its annual trend from 2013 to 2018 (2014 to 2018 at NIMS-YN) was plotted for each site. The mass ratios for each species in PM_{2.5} must be an essential entry in PM_{2.5} source inventory and then it can be directly used to identify the PM_{2.5} emission sources, especially in the field of receptor modeling like source apportionment studies.

When looking over the annual concentration trends of 5 major species at each NIMS shown in Fig. 8, initially SO₄²⁻, OC, and EC decreased sharply at NIMS-JB in a medium-sized urban area after peaking in 2015, but NO₃⁻ increased. However, SO₄²⁻, OC, EC, and NH₄⁺ decreased in 2018 compared to 2013 by 26.2%, 21.4%, 37.6%, and 8.1%, respectively. On the other hand, NO₃⁻ increased by 11.6%. At NIMS-SM in the Seoul metropolitan area, most species showed a decreasing trend during 2013–2015, but there were stagnant or increasing species during 2015–2017. These include SO₄²⁻, NO₃⁻, NH₄⁺, and EC, which decreased in 2018 compared to 2013 by 55.6%, 17.6%, 21.5%, and 33.5%, respectively, but OC increased by 5.0%. At NIMS-HN in a rural/agricultural area, SO₄²⁻, NH₄⁺, and EC decreased in 2018 compared to 2013 by 15.3%, 3.6%, and 19.0%, respectively, whereas NO₃⁻ and OC increased by 33.4% and 0.9%, respectively. For NIMS-BN in a western marine area, SO₄²⁻, NH₄⁺, OC, and EC decreased by 47.1%, 39.9%, 30.0%, and 45.8%, respectively, but NO₃⁻ increased by 8.4%. At NIMS-YN in an industrial area, NH₄⁺, OC, and EC decreased in 2018 compared to 2014 by 4.4%, 8.1%, and 42.8%, respectively, while SO₄²⁻ and

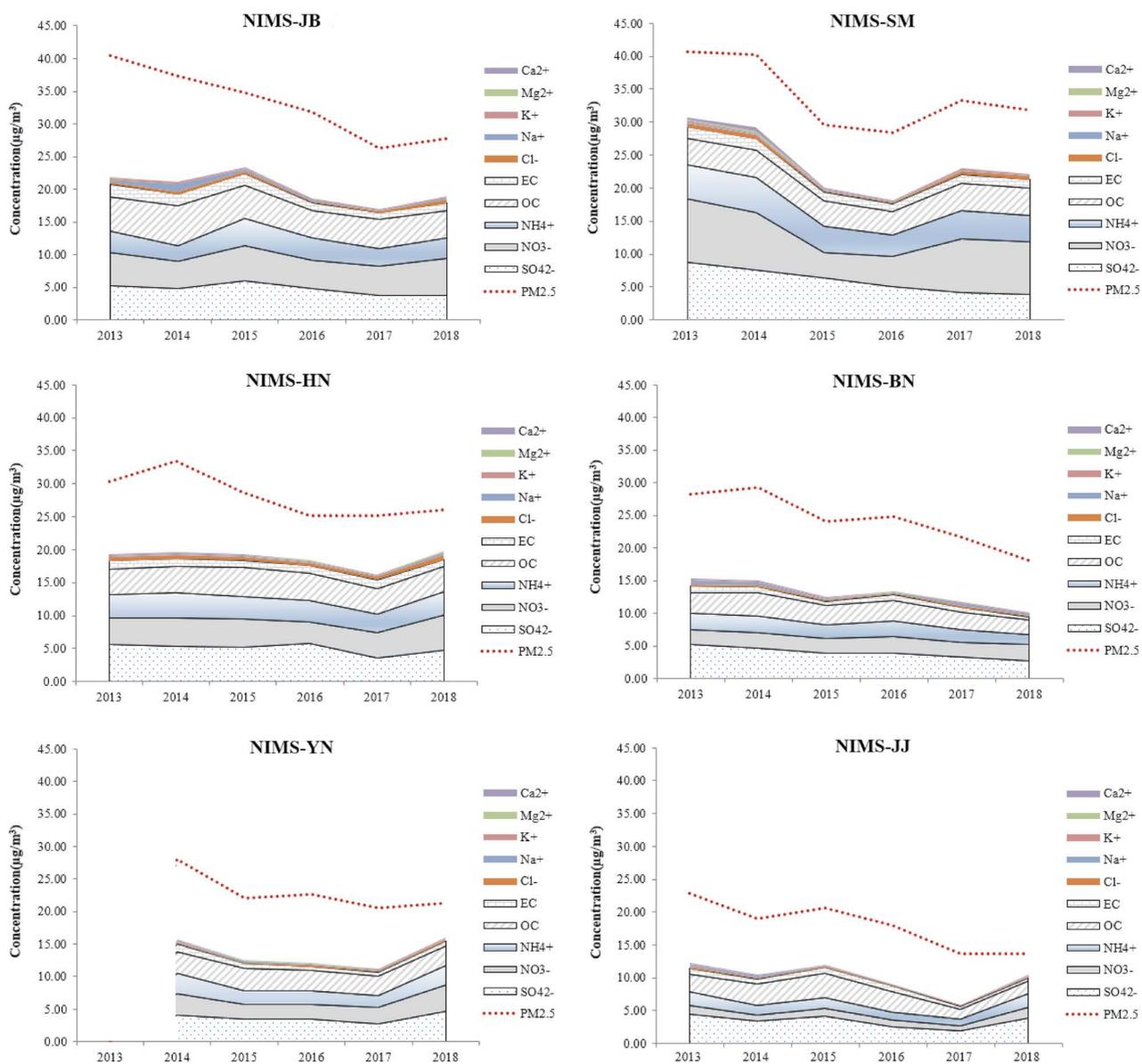


Fig. 7. Trends of annual average concentrations for 10 chemical species in $PM_{2.5}$ (Study-2).

NO_3^- increased by 15.4% and 21.4%, respectively. Lastly, at NIMS-JJ in a southern marine area, SO_4^{2-} , NH_4^+ , OC, and EC decreased in 2018 compared to 2013 by 13.0%, 3.6%, 30.2%, and 44.9%, respectively, while NO_3^- increased by 30.4%. When comparing reduction rate at all 6 NIMSs from 2013 to 2018 (2014 to 2018 at NIMS-YN), EC showed the highest reduction of 37.2%, and SO_4^{2-} , NH_4^+ , and OC showed high reductions of 23.7%, 13.5%, and 14.0%, respectively. On the other hand, only NO_3^- increased by 14.6%.

Trends in terms of mass ratio at each site are presented

in Table S1, which provides all statistics for 5 major species in detail. Briefly, the mass ratio at NIMS-JB showed an increasing trend in 2018 compared to 2013, except for EC. That is, the mass ratio of EC to $PM_{2.5}$ decreased slightly from 0.048 in 2013 to 0.043 in 2018. At NIMS-SM in Seoul, the mass ratio of SO_4^{2-} to $PM_{2.5}$ decreased from 0.214 in 2013 to 0.122 in 2018. However, NO_3^- increased slightly from 0.236 to 0.249, and OC increased from 0.098 to 0.131. At NIMS-HN in an agricultural area, NO_3^- increased from 0.131 to 0.204 and NH_4^+ increased slightly from 0.121 to 0.136. On the other hand, at

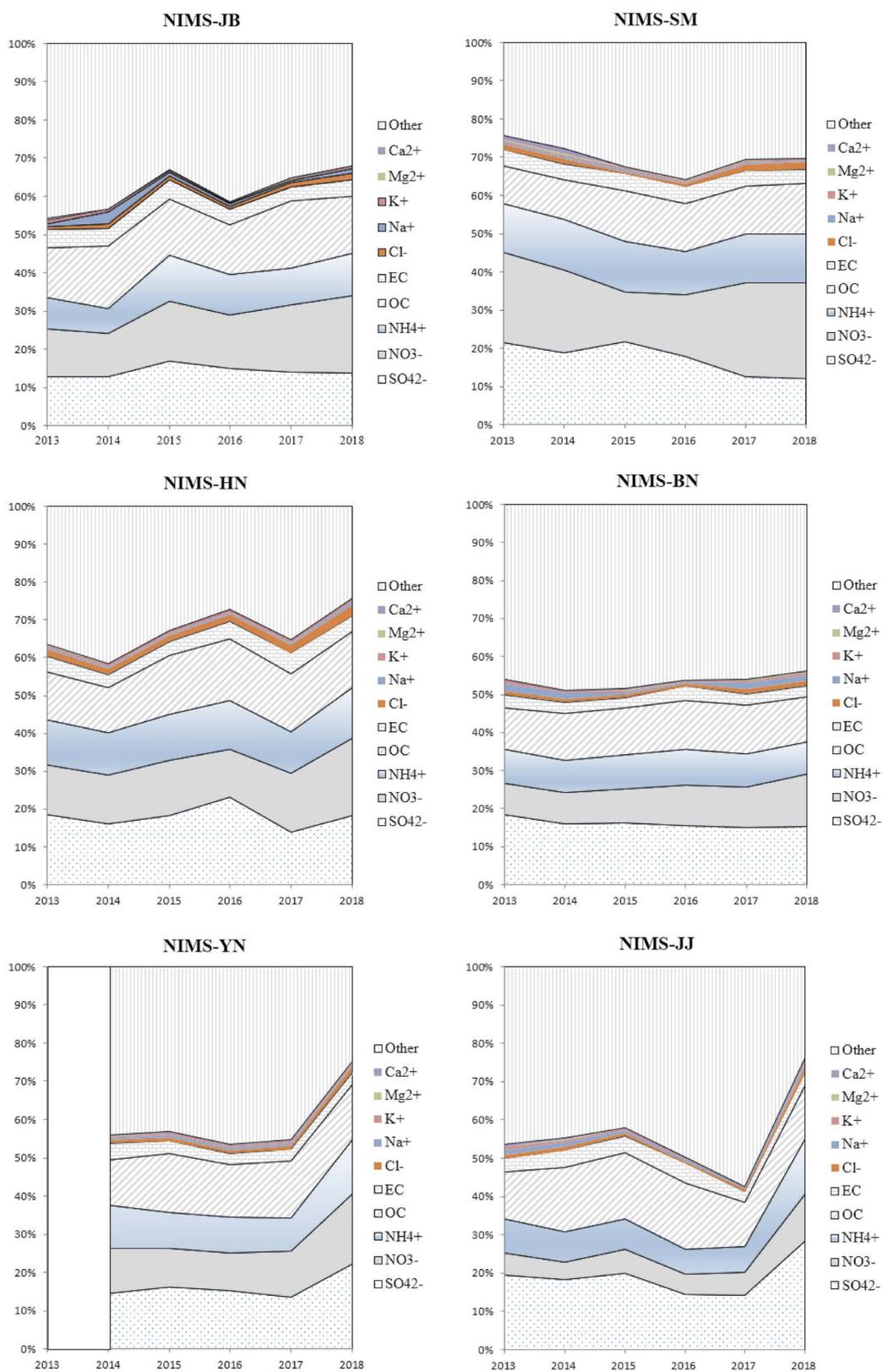


Fig. 8. Stacked plots showing annual trends of 10 chemical compositions as percentage in PM_{2.5} (Study-2).

NIMS-BN on a western island, SO_4^{2-} decreased from 0.185 to 0.153, but NO_3^- greatly increased from 0.081 to 0.137. At NIMS-YN in an industrial area, EC decreased slightly from 0.043 in 2014 to 0.032 in 2018, and SO_4^{2-} and NO_3^- increased greatly from 0.146 to 0.221 and from 0.116 to 0.185, respectively. Lastly, at NIMS-JJ on a southern island, EC decreased slightly from 0.037 in 2013 to 0.034 in 2018, but other species such as SO_4^{2-} (0.195 to 0.283), NO_3^- (0.057 to 0.123), and NH_4^+ (0.090 to 0.144) increased greatly.

3.3 Chemical Characteristics of 8 Ionic Species, EC/OC, and 14 Elements in $\text{PM}_{2.5}$ for Step 3

Mass ratios of each chemical species in $\text{PM}_{2.5}$ is as important as the mass concentration of each species when characterizing $\text{PM}_{2.5}$ physicochemical properties, apportioning quantitative $\text{PM}_{2.5}$ sources, and seeking proper control measures. Among the chemical species measured at 6 NIMSs, samples for a total of 24 chemical species including 8 ionic species (SO_4^{2-} , NO_3^- , NH_4^+ , Cl^- , Na^+ , K^+ , Mg^{2+} , and Ca^{2+}), 2 carbon components (OC, EC), and 14 other elements (Si, Ti, V, Cr, Mn, Fe, Ni, Cu, Zn, As, Se, Br, Ba, and Pb) were analyzed after proper data pretreatment, as described in §2.2. Again, it is noted that the following study is focused on 2018 because the samples for all 14 inorganic elements have been provided at all 6 NIMSs since then.

Table S2 shows a summary of average concentration in unit of $\mu\text{g}/\text{m}^3$ and mass ratio for major chemical species and the other species groups measured at each NIMS in 2018. For the inorganic elements in Table S1, the sum of averaged 14 elements was calculated by $501.5 \text{ ng}/\text{m}^3$ and its mass ratio to $\text{PM}_{2.5}$ ($21.9 \mu\text{g}/\text{m}^3$) was 2.30%. The sum of mass ratios for each element was observed in the range of 1.74% to 2.82% over 6 NIMSs in the order of NIMS-YN (2.82%) > NIMS-JB (2.81%) > NIMS-JJ (2.60%) > NIMS-BN (1.92%) > NIMS-SM (1.90%) > NIMS-HN (1.74%). The elements were generally higher in industrial/urban areas than in agricultural areas. According to Matawie *et al.* (2015), the mass fraction was higher in mobile sources in Raipur, India.

In this study, average fractions for 24 measured species were in the range of 67.1% at NIMS-BN to 89.7% at NIMS-JJ with a mean of 82.7% for all 6 sites. The 3 major elements such as Si, Fe, and Zn were abundant in $\text{PM}_{2.5}$. Average concentrations for all samples were highest for Si ($208.4 \text{ ng}/\text{m}^3$), Fe ($289.3 \text{ ng}/\text{m}^3$), and Zn ($43.1 \text{ ng}/\text{m}^3$), followed sequentially by Pb ($14.8 \text{ ng}/\text{m}^3$) > Mn

($10.4 \text{ ng}/\text{m}^3$) > Br ($7.6 \text{ ng}/\text{m}^3$) > Cu ($5.4 \text{ ng}/\text{m}^3$) > V ($3.6 \text{ ng}/\text{m}^3$). On the other hand, those of the 3 minor elements were as follows: Ba ($2.3 \text{ ng}/\text{m}^3$) > Ni ($1.6 \text{ ng}/\text{m}^3$) > Cr ($1.2 \text{ ng}/\text{m}^3$).

According to a report from the Seoul Research Institute for Health and Environment (SRIHE) (Um *et al.*, 2020), the NO_3^- in Seoul was 24% of the $\text{PM}_{2.5}$ automobile exhaust source. In addition, SO_4^{2-} accounted for 14%, NH_4^+ 12%, OC 16%, and EC 12%. All the other inorganic elements accounted for only 2% of $\text{PM}_{2.5}$. Thus, they reported that all the measured species including ions, EC/OC, and elements accounted for 85% of $\text{PM}_{2.5}$ mass. As $\text{PM}_{2.5}$ mass increased, the mass ratio of ion species showed a tendency to increase in general.

Carbon components are usually classified as OC and EC. Since OC detected only carbon by our analytical method, some elements such as O, H, and S bound to carbon were excluded during component analysis. When estimating organic matter (OM), a reasonable constant is often multiplied by OC to make up for unmeasured chemicals in $\text{PM}_{2.5}$. Um *et al.* (2020) used 1.8 as an appropriate constant in Seoul city, and then they estimated carbon component such that $\text{OM} = 1.8 \times \text{OC}$, and then TC (total carbon) = $\text{OM} + \text{EC} = 1.8 \times \text{OC} + \text{EC}$. After using the constant in 2018, a result showed that TC in Seoul accounted for 32% of $\text{PM}_{2.5}$ compared to the initial 28% (i.e., 16% of OC plus 12% of EC). Fig. 9 shows average chemical compositions in $\text{PM}_{2.5}$ at each NIMS for 24 chemical species after converting OM into $1.8 \times \text{OC}$ from data monitored hourly in 2018.

On the other hand, Chen and Yu (2007) reported that a value of 2.3 ± 0.3 was applicable to convert OC to OM in Hong Kong. A OM/OC ratio of 2.2 was applicable to continental air mass, and 1.9 was applicable to marine air mass. However, Cheng *et al.* (2016) used 1.4 in many global megacities, and they used the equation $\text{OM} = 1.4 \times \text{OC}$. They also estimated a soil component such that $\text{Soil} = 2.2\text{Al} + 2.49\text{Si} + 1.63\text{Ca} + 2.42\text{Fe} + 1.94\text{Ti}$. Further, they estimated the unidentified chemical portion of $\text{PM}_{2.5}$ such that unidentified portion = $\text{PM}_{2.5} - \text{OM} - \text{EC} - \text{SNA} - \text{Soil}$, where SNA stands for SO_4^{2-} , NO_3^- , and NH_4^+ . They noted that the average chemical composition of $\text{PM}_{2.5}$ in 38 mega cities around the world in 2013 was: OM of $30 \pm 8\%$, EC of $7 \pm 4\%$, SNA of $36 \pm 10\%$, and soil $10 \pm 5\%$. Unidentified fractions were $19 \pm 12\%$. In their study, the unidentified (i.e., unmeasured) fractions of major cities were LA 11% ($\text{PM}_{2.5}$ $16.2 \mu\text{g}/\text{m}^3$), New York 14% ($\text{PM}_{2.5}$ $9.1 \mu\text{g}/\text{m}^3$), London 41%

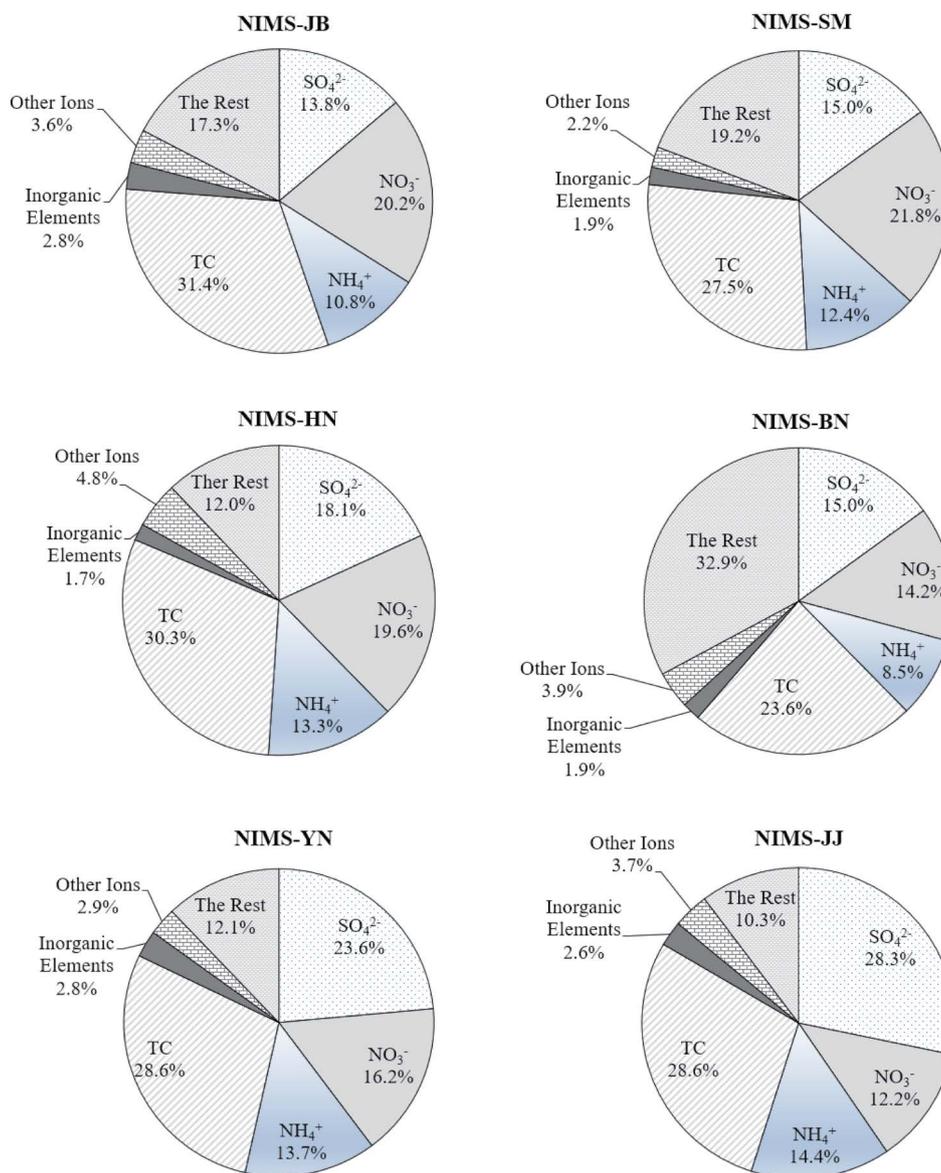


Fig. 9. Average chemical compositions in PM_{2.5} for 24 chemical species.

(PM_{2.5} 22.4 μg/m³), Paris 20% (PM_{2.5} 19.8 μg/m³), Beijing 18% (PM_{2.5} 87.0 μg/m³), Tianjin 28% (PM_{2.5} 95.7 μg/m³), X'ian 15% (PM_{2.5} 102.2 μg/m³), and Delhi 18% (PM_{2.5} 143.0 μg/m³). According to the results of other studies, the unidentified fractions in PM_{2.5} were 5% in Beijing and 9.7% in Tianjin from 2014 to 2015 (Huang *et al.*, 2017), and 14.6% in Luoyang from 2019 to 2020 (Xu *et al.*, 2021), Qatar's Doha from May to Dec 2015 was estimated to be 20.3% (Javed *et al.*, 2021).

Back to our study in Korea, we multiplied OC by a constant of 1.8 as suggested by Um *et al.* (2020). The

average TC at all 6 sites accounted for 28.3% of PM_{2.5} with a range of 23.6% at NIMS-BN to 31.4% at NIMS-JB, as shown in Table S2. Thus, the TC in a medium-sized urban area was much higher than those in a marine area and even in large urban and industrial areas. It seemed that the latter areas were better controlled than the former area in terms of combustion activities of fossil fuels. In addition, the measured fractions of PM_{2.5} for 24 species were in the range of 67.1% at NIMS-BN to 89.7% at NIMS-JJ with a mean of 82.7% for all 6 sites. Both sites showing minimum and maximum values were

in the same marine environment even though both sites showed much lower $PM_{2.5}$ levels than inland sites. We wondered what kinds of unidentified components accounted for 32.9% of $PM_{2.5}$ mass at NIMS-BN, far exceeding the other 5 sites' average of 14.2%. According to Almeida *et al.* (2006), the unmeasured fraction in $PM_{2.5}$ might be partly due to the presence of water associated with PM and errors in the estimation of chemical species. In addition, Tsai and Kuo (2005) reported that water content in $PM_{2.5}$ was higher in coastal areas than in urban areas. Further, it was higher in spring than in winter and higher at night than in the daytime. Many other earlier studies focused on water content and hygroscopic growth of PM under ambient conditions (Ueda *et al.*, 2000; Meng *et al.*, 1995a). Generally, when RH was high, hygroscopic components like NH_4NO_3 in PM took up water from the atmosphere so that particles might gain mass several times their dry weight, and then $PM_{2.5}$ might contain water as an unidentified component under very humid conditions (Kajino *et al.*, 2006). Perrino *et al.* (2016) insisted that water content in PM was roughly proportional to the soil content from long-range transported desert dust (about 5%), but dust did not contribute to water content. They also argued that the water content under winter stability conditions was dependent on NH_4NO_3 and constituted up to 22% of the total PM_{10} .

For reference during the study year 2018, the NIMS-BN on site had an annual average F/C ratio of 0.51, a temperature of 6.1°C, precipitation of 831.1 mm/yr, a WS of 4.3 m/s, a WD of 227.0°, and a RH of 67.9%. However, the NIMS-JJ site had a F/C ratio of 0.47, a temperature of 14.5°C, precipitation of 1,345.8 mm/yr, WS of 3.1 m/s, WD of 194.4°, and RH of 68.6%. Also, Asian yellow dust events occurred 7 times in 2018 at NIMS-BN, but none occurred at NIMS-JJ (KMA, 2021). The NIMS-BN environment was colder, showed much less precipitation, and much faster WS than the other 5 sites. Thus, probable reasons for the smallest unidentified fraction at NIMS-BN were: 1) more water content due to a humid marine environment as well as frequent Asian dust occurrences, 2) colder weather condition below zero to facilitate water binding and building up more hygroscopic NH_4NO_3 in winter, and 3) unaccounted species from China and nearby North Korea with fastest WS compared to the other sites.

4. CONCLUSIONS

The chemical compositions and their mass ratios in $PM_{2.5}$ are particularly important in fundamental studies on identifying and determining $PM_{2.5}$ emission sources. For this study, the PM was monitored hourly for 6 years at 6 different environmental areas (urban/agricultural/industrial/marine areas) in Korea. $PM_{2.5}$ levels in Korea far exceeded the 2021 WHO guidelines and slightly exceeded the 2018 Korean NAAQS. When classifying $PM_{2.5}$ samples into 5 classes of air quality conditions, we observed that 45.0% of all datasets were assorted in the good class. It was found that PM_{10} and $PM_{2.5}$ were negatively correlated with WS; however, PM was influenced by wind direction rather than wind speed in certain areas. Since annual $PM_{2.5}$ concentration steadily decreased nationwide, each species was also decreased. One of the reasons for decreasing PM_{10} and $PM_{2.5}$ might be caused by changes in citizens' awareness of PM due to enforcement of various laws and policies in Korea. In addition, it was found that the $PM_{2.5}$ concentration in China decreased by more than 50% during the same period, which is considered to be an important external factor in the reduction of $PM_{2.5}$ in Korea. The chemical mass ratio of $PM_{2.5}$ significantly changed based on area. For example, at NIMS-SM in the metropolitan area, the mass ratio of SO_4^{2-} decreased from 0.214 in 2013 to 0.122 in 2018. However, NO_3^- increased slightly from 0.236 to 0.249, and OC increased from 0.098 to 0.131. At NIMS-HN in an agricultural area from 2013 to 2018, NO_3^- increased from 0.131 to 0.204, NH_4^+ and OC increased slightly from 0.127 to 0.136. On the other hand, at NIMS-BN in the western marine area, SO_4^{2-} decreased from 0.185 to 0.153, and NO_3^- greatly increased from 0.081 to 0.137. The elements were higher in industrial/urban areas than in agricultural areas. In addition, the average TC for 6 sites was 28.3% of $PM_{2.5}$. The TC in a medium-sized urban area was much higher than those in marine areas or even in large populated urban areas and industrial areas. It seemed to be that the latter areas were better controlled than the former area in terms of combustion activities of fossil fuels.

The chemical species in $PM_{2.5}$ such as various ions, carbons, and elements provide fundamental information to help qualitatively identify emission sources, quantitatively determining the sources, suitably managing PM action plans, and reasonably establishing government policies. For the efficient control and management of air

pollutants and to establish a reasonable air environment policy, it is necessary to present and implement fine dust management measures and countermeasures that match the circumstances of each local government. It is considered that the concentration trend analysis of PM_{2.5} and chemical components in six regions in Korea can serve as an important basic data for each region's air pollutant emission reduction and management plan.

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SUPPLEMENTARY MATERIALS

Table S1. Overall summary of annual average concentrations ($\mu\text{g}/\text{m}^3$) and mass ratios for 5 major chemical species (Study-2).

Site	Year	PM _{2.5}	SO ₄ ²⁻	NO ₃ ⁻	NH ₄ ⁺	OC	EC	SNA ^{a)}	OC+EC	SNA+OC+EC	Avg (SNA)	Avg (OC+EC)	Avg (SNA+EC+OC)
NIMS-JB	2013	40.49 (1.00)	5.180 (0.128)	5.065 (0.125)	3.293 (0.081)	5.326 (0.132)	1.922 (0.047)	13.538 (0.334)	7.248 (0.179)	20.786 (0.513)	12.728 (0.391)	6.381 (0.194)	19.109 (0.584)
	2014	37.28 (1.00)	4.762 (0.128)	4.225 (0.113)	2.393 (0.064)	6.121 (0.164)	1.737 (0.047)	11.380 (0.305)	7.858 (0.211)	19.238 (0.516)			
	2015	34.82 (1.00)	5.921 (0.170)	5.407 (0.155)	4.201 (0.121)	5.115 (0.147)	1.735 (0.050)	15.528 (0.446)	6.850 (0.197)	22.378 (0.643)			
	2016	31.78 (1.00)	4.785 (0.151)	4.388 (0.138)	3.406 (0.107)	4.092 (0.129)	1.301 (0.041)	12.580 (0.396)	5.393 (0.170)	17.972 (0.566)			
	2017	26.23 (1.00)	3.680 (0.140)	4.603 (0.175)	2.560 (0.098)	4.556 (0.174)	0.996 (0.038)	10.843 (0.413)	5.552 (0.212)	16.395 (0.625)			
	2018	27.80 (1.00)	3.823 (0.138)	5.653 (0.203)	3.025 (0.109)	4.185 (0.151)	1.200 (0.043)	12.501 (0.450)	5.385 (0.194)	17.886 (0.643)			
NIMS-SM	2013	40.67 (1.00)	8.697 (0.214)	9.606 (0.236)	5.226 (0.129)	3.964 (0.097)	1.796 (0.044)	23.529 (0.579)	5.760 (0.142)	29.289 (0.720)	17.426 (0.507)	5.441 (0.162)	22.868 (0.669)
	2014	40.29 (1.00)	7.564 (0.18)8	8.755 (0.217)	5.285 (0.131)	4.208 (0.104)	1.704 (0.042)	21.603 (0.536)	5.912 (0.147)	27.516 (0.683)			
	2015	29.58 (1.00)	6.436 (0.218)	3.854 (0.130)	3.865 (0.131)	3.954 (0.134)	1.334 (0.045)	14.155 (0.478)	5.288 (0.179)	19.443 (0.657)			
	2016	28.34 (1.00)	5.089 (0.180)	4.553 (0.161)	3.220 (0.114)	3.561 (0.126)	1.239 (0.044)	12.862 (0.454)	4.799 (0.169)	17.662 (0.623)			
	2017	33.22 (1.00)	4.171 (0.126)	8.183 (0.246)	4.175 (0.12)6	4.230 (0.127)	1.298 (0.039)	16.529 (0.498)	5.528 (0.166)	22.058 (0.664)			
	2018	31.78 (1.00)	3.861 (0.122)	7.915 (0.249)	4.104 (0.129)	4.164 (0.131)	1.195 (0.038)	15.880 (0.500)	5.358 (0.169)	21.238 (0.668)			
NIMS-HN	2013	30.35 (1.00)	5.588 (0.184)	3.985 (0.131)	3.663 (0.121)	3.823 (0.126)	1.284 (0.042)	13.236 (0.436)	5.108 (0.168)	18.344 (0.604)	12.600 (0.450)	5.205 (0.187)	17.805 (0.637)
	2014	33.48 (1.00)	5.371 (0.160)	4.302 (0.128)	3.757 (0.112)	4.037 (0.121)	1.123 (0.034)	13.430 (0.401)	5.160 (0.154)	18.590 (0.555)			
	2015	28.66 (1.00)	5.234 (0.183)	4.197 (0.146)	3.464 (0.121)	4.492 (0.157)	1.039 (0.036)	12.895 (0.450)	5.532 (0.193)	18.426 (0.643)			
	2016	25.23 (1.00)	5.825 (0.231)	3.170 (0.126)	3.309 (0.131)	4.075 (0.162)	1.174 (0.047)	12.303 (0.488)	5.249 (0.208)	17.552 (0.696)			
	2017	25.18 (1.00)	3.521 (0.140)	3.909 (0.155)	2.723 (0.108)	3.908 (0.155)	1.376 (0.055)	10.153 (0.403)	5.285 (0.210)	15.437 (0.613)			
	2018	26.07 (1.00)	4.734 (0.182)	5.314 (0.204)	3.532 (0.136)	3.860 (0.148)	1.041 (0.040)	13.581 (0.521)	4.900 (0.188)	18.481 (0.709)			

Table S1. Continued.

Site	Year	PM _{2.5}	SO ₄ ²⁻	NO ₃ ⁻	NH ₄ ⁺	OC	EC	SNA ^{a)}	OC + EC	SNA + OC + EC	Avg (SNA)	Avg (OC + EC)	Avg (SNA + OC + EC)
NIMS-BN	2013	28.28 (1.00)	5.224 (0.185)	2.277 (0.081)	2.547 (0.090)	3.066 (0.108)	1.023 (0.036)	10.048 (0.355)	4.089 (0.145)	14.137 (0.500)	8.483 (0.350)	3.738 (0.154)	12.221 (0.503)
	2014	29.25 (1.00)	4.705 (0.161)	2.408 (0.082)	2.456 (0.084)	3.586 (0.123)	0.885 (0.030)	9.569 (0.327)	4.471 (0.153)	14.040 (0.480)			
	2015	24.12 (1.00)	3.937 (0.163)	2.154 (0.089)	2.146 (0.089)	2.995 (0.124)	0.622 (0.026)	8.237 (0.342)	3.618 (0.150)	11.855 (0.492)			
	2016	24.77 (1.00)	3.853 (0.15)6	2.635 (0.106)	2.330 (0.094)	3.188 (0.129)	0.927 (0.037)	8.819 (0.356)	4.115 (0.166)	12.934 (0.522)			
	2017	21.74 (1.00)	3.257 (0.150)	2.341 (0.108)	1.869 (0.086)	2.783 (0.128)	0.652 (0.030)	7.466 (0.343)	3.435 (0.158)	10.902 (0.501)			
	2018	18.05 (1.00)	2.761 (0.153)	2.468 (0.137)	1.530 (0.085)	2.148 (0.119)	0.554 (0.031)	6.760 (0.374)	2.702 (0.150)	9.462 (0.524)			
NIMS-YN	2014	27.98 (1.00)	4.083 (0.146)	3.255 (0.116)	3.151 (0.113)	3.344 (0.120)	1.209 (0.043)	10.489 (0.375)	4.552 (0.163)	15.041 (0.538)	8.965 (0.393)	3.979 (0.175)	12.944 (0.568)
	2015	22.01 (1.00)	3.539 (0.16)1	2.246 (0.102)	2.053 (0.093)	3.404 (0.155)	0.733 (0.033)	7.837 (0.356)	4.138 (0.188)	11.975 (0.544)			
	2016	22.62 (1.00)	3.411 (0.151)	2.275 (0.101)	2.134 (0.094)	3.106 (0.137)	0.631 (0.028)	7.819 (0.346)	3.737 (0.165)	11.556 (0.511)			
	2017	20.49 (1.00)	2.765 (0.135)	2.492 (0.122)	1.746 (0.085)	3.068 (0.150)	0.637 (0.031)	7.003 (0.342)	3.706 (0.181)	10.709 (0.523)			
	2018	21.35 (1.00)	4.711 (0.221)	3.953 (0.185)	3.011 (0.141)	3.072 (0.144)	0.692 (0.032)	11.675 (0.547)	3.764 (0.176)	15.439 (0.723)			
NIMS-JJ	2013	22.89 (1.00)	4.471 (0.195)	1.297 (0.057)	2.054 (0.090)	2.772 (0.121)	0.847 (0.037)	7.822 (0.342)	3.619 (0.158)	11.441 (0.500)	6.113 (0.345)	3.428 (0.189)	9.541 (0.534)
	2014	18.95 (1.00)	3.445 (0.182)	0.906 (0.048)	1.476 (0.078)	3.200 (0.169)	0.836 (0.044)	5.827 (0.308)	4.037 (0.213)	9.863 (0.521)			
	2015	20.70 (1.00)	4.136 (0.200)	1.273 (0.062)	1.651 (0.080)	3.595 (0.17)4	0.904 (0.044)	7.060 (0.341)	4.499 (0.217)	11.559 (0.558)			
	2016	17.97 (1.00)	2.580 (0.144)	0.973 (0.054)	1.155 (0.064)	3.102 (0.173)	0.949 (0.053)	4.708 (0.262)	4.051 (0.225)	8.759 (0.487)			
	2017	13.70 (1.00)	1.936 (0.141)	0.822 (0.060)	0.943 (0.069)	1.580 (0.115)	0.381 (0.028)	3.700 (0.270)	1.961 (0.143)	5.661 (0.413)			
	2018	13.74 (1.00)	3.888 (0.283)	1.691 (0.123)	1.981 (0.144)	1.934 (0.141)	0.467 (0.034)	7.560 (0.550)	2.401 (0.175)	9.961 (0.725)			

^{a)}SNA = SO₄²⁻ + NO₃⁻ + NH₄⁺

Table S2. A summary of average concentrations ($\mu\text{g}/\text{m}^3$) and mass ratios for major species, Total Carbon, 5 other ions and 14 elements (Study-3).

Site	PM _{2.5}	SO ₄ ²⁻	NO ₃ ⁻	NH ₄ ⁺	Other ions ^{a)}	TC ^{b)}	Elements ^{c)}	The rest ^{d)}	Total
NIMS-JB	27.6	3.800 (0.138)	5.571 (0.202)	2.984 (0.108)	0.991 (0.036)	8.669 (0.314)	0.776 (0.028)	4.776 (0.173)	22.790 (0.827)
NIMS-SM	24.2	3.622 (0.150)	5.278 (0.218)	3.007 (0.124)	0.536 (0.022)	6.664 (0.275)	0.460 (0.019)	4.648 (0.192)	19.567 (0.808)
NIMS-HN	25.3	4.598 (0.181)	4.968 (0.196)	3.379 (0.133)	1.225 (0.048)	7.687 (0.303)	0.442 (0.017)	3.047 (0.120)	22.298 (0.880)
NIMS-BN	19.1	2.864 (0.150)	2.724 (0.142)	1.633 (0.085)	0.748 (0.039)	4.515 (0.236)	0.367 (0.019)	6.297 (0.329)	12.851 (0.671)
NIMS-YN	21.4	5.030 (0.236)	3.461 (0.162)	2.933 (0.137)	0.622 (0.029)	6.116 (0.286)	0.602 (0.028)	2.588 (0.121)	18.764 (0.879)
NIMS-JJ	14.0	3.949 (0.283)	1.709 (0.122)	2.013 (0.144)	0.517 (0.037)	3.991 (0.286)	0.363 (0.026)	1.433 (0.103)	12.542 (0.897)
Average	21.9	3.977 (0.189)	3.952 (0.174)	2.658 (0.122)	0.773 (0.035)	6.274 (0.283)	0.501 (0.023)	3.798 (0.173)	18.135 (0.827)

^{a)}Other ions are Cl⁻, Na⁺, K⁺, Mg²⁺, and Ca²⁺;

^{b)}TC (total carbon) = OM (organic matter) + EC, where OM = 1.8 × OC;

^{c)}Element stands for Si, Ti, V, Cr, Mn, Fe, Ni, Cu, Zn, As, Se, Br, Ba, and Pb;

^{d)}The rest = Unidentified species = PM_{2.5} - (SO₄²⁻ + NO₃⁻ + NH₄⁺ + Other ions + TC + Elements)