

1 **Evaluation of regional background particulate matter**
2 **concentration based on vertical distribution**
3 **characteristics**

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1 **Abstract**

2 Heavy regional particulate matter (PM) pollution in China has resulted in an
3 important and urgent need for joint control actions among cities. It's advisable to
4 improve the understanding of regional background concentration of PM for the
5 development of efficient and effective joint control policies. With the increase of
6 height the influence of source emission on local air quality decreases with altitude, but
7 the characteristics of regional pollution gradually become obvious. A method to
8 estimate regional background PM concentration is proposed in this paper, based on
9 the vertical characteristics of periodic variation in the atmospheric boundary layer
10 structure and particle mass concentration, as well as the vertical distribution of
11 particle size, chemical composition and pollution source apportionment. According to
12 the method, the averaged regional background PM_{2.5} concentration in July, August
13 and September 2009, being extracted from the original time series in Tianjin, was 40
14 $\pm 20 \mu\text{g m}^{-3}$, $64 \pm 17 \mu\text{g m}^{-3}$ and $53 \pm 11 \mu\text{g m}^{-3}$, respectively.

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1 **1 Introduction**

2 Atmospheric particulate matter (PM) has drawn considerable attention because it has
3 been associated with many urban environmental problems, such as acid precipitation,
4 decreasing visibility and climate change (Zeng and Hopke, 1989; Charlson et al.,
5 1992; Schwartz et al., 1996; Chameides et al., 1999). PM has also been implicated in
6 human mortality and morbidity (Dockery et al., 1993; Tie et al., 2009; Lagudu et al.,
7 2011). Among the various sizes of atmospheric PM, PM_{2.5} (PM with aerodynamic
8 diameter less than 2.5 μm) is considered to be of great significance due to its links to
9 human respiratory health (Englert, 2004) , regional-scale air pollution (Husar et al.,
10 1981; Chameides et al., 1999) ,and potential acid rain enhancement (Cao et al.2013).

11 The combination of rapid industrialization and urbanization has resulted in
12 considerable environmental problems throughout China, especially in the clusters of
13 cities (Shao et al., 2006). The coexistence of numerous air pollutants with high
14 concentrations and the complicated interactions among them leads to the formation of
15 an air pollution complex(Shao et al., 2006; Zhu et al., 2011). One of the major
16 pollutants is PM(Tie et al., 2006; Liu et al., 2011; Chen et al., 2012; Han et al., 2013).
17 The origin of PM is complex. It involves both primary emissions as well as secondary
18 particle production due to chemical reactions in the atmosphere (Shi et al., 2011; Tian
19 et al., 2013; Hu et al., 2013; Guo et al., 2013). With a lifetime of days to weeks in the
20 lower atmosphere, PM_{2.5} can be transported thousands of kilometers (Hagler et al.,
21 2006). The trans-boundary transport of PM_{2.5} and the gaseous precursors has
22 significant influence on the regional background PM level in the cluster of cities. In
23 order to study the regional-scale PM pollution and develop efficient joint control
24 policies, it's necessary to improve understanding of regional background PM
25 concentration.

26 Background concentration has been defined as concentration observed at a site “that is
27 not affected by local sources of pollution” (WHO, 1980; Menichini et al., 2007).
28 McKendry (2006) defined background concentration as one of “those pollutants
29 arising from local natural processes together with those transported into an airshed

1 from afar (the latter may be either natural or anthropogenic in origin)". Background
2 concentration in this paper is defined to include collective contributions from regional
3 anthropogenic and natural emissions and long-range transport.

4 Background concentrations are not constant because of meteorological variability,
5 complexity of chemical reactions, as well as spatially and temporally varying
6 emissions. Regional-scale PM pollution is associated with synoptic scenarios that
7 induce the transfer, accumulation and the formation of pollutants at regional scales
8 (Ronald et al., 2007). Simply taking measurements at local scales is not well suited to
9 adequately investigate the regional background concentration. There is always the
10 possibility that the "air quality background monitoring station" is directly influenced
11 by local emission sources and thus not truly representative of the background level
12 (Tchepele et al., 2010). That is to say, background concentration can hardly be
13 measured directly, so it is critical to choose representative and appropriate values.
14 Usually, by setting some restrictions to identify and remove the influence of local
15 pollution, background concentration can be determined indirectly. There are several
16 studies mentioning the methods for determining the background concentration. These
17 methods can be classified into 4 categories. (1) The physical methods identify the
18 regional pollution process and local pollution process via synoptic situation, duration
19 of the synoptic system, consistency of vertical wind, and atmospheric stability,
20 particle size distribution, etc., and then the data of the "background period" influenced
21 by regional processes are selected (Pérez et al., 2008). (2) The chemical methods
22 identify the regional process according to chemical composition in PM and
23 synchronous observation of other pollutants, and then remove the data influenced by
24 local processes (Menichini, 2007). (3) The statistical methods use discriminant
25 analysis, cluster analysis and principal component analysis (PCA) to identify the data
26 that characterize the regional background PM (Langford et al., 2009; Tchepele et al.,
27 2010). (4) Numerical simulation methods use trajectory models and atmospheric
28 dynamics-chemical coupled models to simulate the regional background pollution
29 (Dreyer et al., 2009, Tchepele et al. 2010).

1 With the increase of height, the influence of source emission on local air quality
2 decreases with altitude, but the characteristics of regional pollution gradually become
3 obvious. Influenced by atmospheric dynamics and thermal effects, meteorological
4 variables and pollutant measurements at different heights within the boundary layer
5 could represent different horizontal scales of pollution. Sites at near ground height
6 (5-10m) are influenced extensively by human activities, and the data observed at these
7 sites could represent the street scale. Impacts from local disturbance weakening with
8 height gradually and observations at greater heights could represent larger horizontal
9 scales. When the height increases to the top of the urban atmospheric boundary layer,
10 observations can represent urban scales. Heights above the urban boundary layer
11 could to some extent reflect the characteristics of regional scales. Tall tower is
12 commonly used in observation of boundary layer meteorological,
13 micrometeorological and atmospheric chemical variables, e.g. vertical profile and
14 fluxes(Heintzenberg et al., 2008; Brown et al., 2013; Heintzenberg et al., 2013;
15 Andreae et al.,2015). The footprint concept is capable of linking observed data
16 collected at the different height levels of tower to spatial context. The integral beneath
17 the foot-print function expresses the total surface influence on the signal measured by
18 the sensor at height above the surface(Schmid, 2002; Ding et al., 2005; Foken et al.,
19 2008). Three main factors affect the size and shape of flux footprint: increase in
20 measurement height, decrease in surface roughness, and change in atmospheric
21 stability from unstable to stable would lead to an increase in size of the footprint
22 (https://en.wikipedia.org/wiki/Flux_footprint).. Combined informations from
23 meteorological data and simultaneous aerosol measurements at the different levels of
24 the tower have allowed to gain insights into transport of aerosols and their vertical
25 distributions strongly depends on meteorological conditions, boundary layer dynamics
26 and physiochemical processes(Guinot, et al., 2006; Pal, et al., 2014). In this paper, the
27 periodic variation in the atmospheric boundary layer structure and PM mass
28 concentrations, as well as the vertical distribution characteristics of particle size,
29 chemical composition and pollution sources were studied to characterize the regional
30 pollution contribution. And on this basis, the height above which influenced relatively

1 less by local pollution emission can be determined and the regional background PM
2 concentration can be extracted from the observation data and estimate by
3 mathematical methods.

4 **2 Data sources and treatment**

5 **2.1 Observation site**

6 The data used in this study were collected at a 255 m meteorological tower which is
7 located at the atmospheric boundary layer observation station(WMO Id.No. 54517,
8 39°04'29.4"N, 117°12'20.1"E) in Tianjin, China, where is a residential and traffic
9 mixing area. There are no industrial pollution sources near the site. Tianjin is adjacent
10 to the BoHai Sea and situated in the eastern part of the Beijing-Tianjin-Hebei area,
11 one of the most heavily polluted areas in China. Tianjin covers an area of 11,300 km²
12 and has a population of 8 million. Due to rapid industrialization and urbanization in
13 recent years, air pollution has become a serious problem in this city.

14 **2.2 Observation method and data treatment**

15 Horizontal wind speed, wind direction, and temperature were measured at 15
16 platforms (5,10, 20, 30, 40, 60, 80, 100, 120, 140, 160, 180, 200, 220 and 250 m)
17 every 10 s and averaged hourly. Three dimensional ultrasonic anemometers
18 (CAST-3D) were mounted at 40 m, 120 m and 220 m to measure the turbulent fluxes.
19 Hourly meteorological data(WMO Id.No. 54517) in the year of 2009 were used in this
20 paper.

21 Mass concentrations of PM_{2.5} were measured using ambient particulate monitor
22 chemiluminescence (TEOMR-RP1400a) at four different heights (2, 40, 120, and 220
23 m) from July 1 to September 30, 2009. The monitor's data output consists of 1-hour
24 and 24-hour average mass concentration updated every 10 minutes and on the
25 hour ,with the precision of $\pm 1.5\mu\text{g m}^{-3}$ (1-hour ave) and $\pm 0.5\mu\text{g m}^{-3}$ (24-hour ave)
26 respectively. Accuracy for mass measurement is $\pm 0.75\%$.

27 In order to study the vertical characteristics of PM chemical composition and sources,

1 twenty-four hour PM₁₀ filter samples were collected from local Beijing time 08:00 to
2 07:00 the next day using medium-volume PM₁₀ samplers (TH-150, Wuhan Tianhong
3 Intelligence Instrumentation Facility) at the heights of 10 m, 40 m, 120 m, and 220 m
4 from August 24 to September 12, 2009. The sampler has a system of automatic
5 constant-flow control. Flow rate of sampling in this study is 100 L min⁻¹, and the
6 relative error of flow is less than 3%. At each height, PM₁₀ filter samplings were
7 equipped with two samplers in parallel: one is for chemical analysis of inorganic
8 composition on polypropylene filters (90 mm in diameter, Beijing Synthetic Fiber
9 Research Institute, China) and the other is for organic composition analyses on
10 quartz-fiber filters (90 mm in diameter, 2500QAT-UP, Pall Life Sciences).

11 Before and after sampling, filters were conditioned for 48 h in darkened desiccators
12 prior to gravimetric determination. The filters were weighed on a electronic
13 microbalance (AX205, Mettler-Toledo, LLC, with a ±0.01 mg sensitivity) in a clean
14 room under constant temperature (20±1 °C) and RH (40±3%). Samples were stored
15 air-tight in a refrigerator at about 4 °C before chemical analyses.

16 Elements (Si, Ti, Al, Mn, Ca, Mg, Na, K, Cu, Zn, Pb, Cr, Ni, Co, Fe, and V) were
17 analyzed by Inductively Coupled Plasma-atomic emission spectroscopy (ICP
18 9000(N+M) Thermo Electron Corporation, USA). Blank filters were processed
19 simultaneously with sample filters. Ultrapure water, both unfiltered and filtered, and
20 nitric acid were also analyzed. The average element values in the blanks were
21 subtracted from those obtained for each sample filter. 10 percent of total samples were
22 analyzed in duplicate to verify sample homogeneity. The precision and accuracy were
23 checked by analysis of an intermediate calibration solution. Extraction efficiencies
24 were evaluated by analysis of the certified reference material from National Research
25 Center of CRM. The recovery value was between 85% and 110%. Calibration check
26 was performed to ensure a relative error no more than 2% for major elements and 5%
27 for trace elements.

28 Water-soluble ions (NH₄⁺, Cl⁻, NO₃⁻, and SO₄²⁻) were analyzed by ion chromatography
29 (DX-120, Dionex Ltd., USA) after extraction by deionized water. External calibration

1 was employed to quantify the ions concentrations. A calibration check with external
2 standards was performed to ensure a relative error no more than 10%. The uncertainty
3 contributions of the calibration curve, calibration solution, and repetitive
4 measurement for unknown sample were taken into account. The expanded uncertainty
5 was 3.8% with a coverage factor $k=2$.

6 The thermal optical carbon analyzer (Desert Research Institute (DRI) Model 2001,
7 Atmoslytic Inc., Calabasas, CA, USA) was used to measure organic carbon (OC) and
8 elemental carbon (EC). The heating process can be found in IMPROVE_A protocol
9 (Chow et al., 2010, 2011; Cao et al., 2003). Field blank and lab blank were considered
10 and all sampling concentrations were revised by blank concentration. The uncertainty
11 contributions of the calibration curve, calibration solution, and repetitive
12 measurement for unknown sample were taken into account. The expanded uncertainty
13 was 7.6% with a coverage factor $k=2$.

14 **3 Vertical variation characteristics of urban boundary structure**

15 **3.1 Thermal and dynamic characteristics in surface layer**

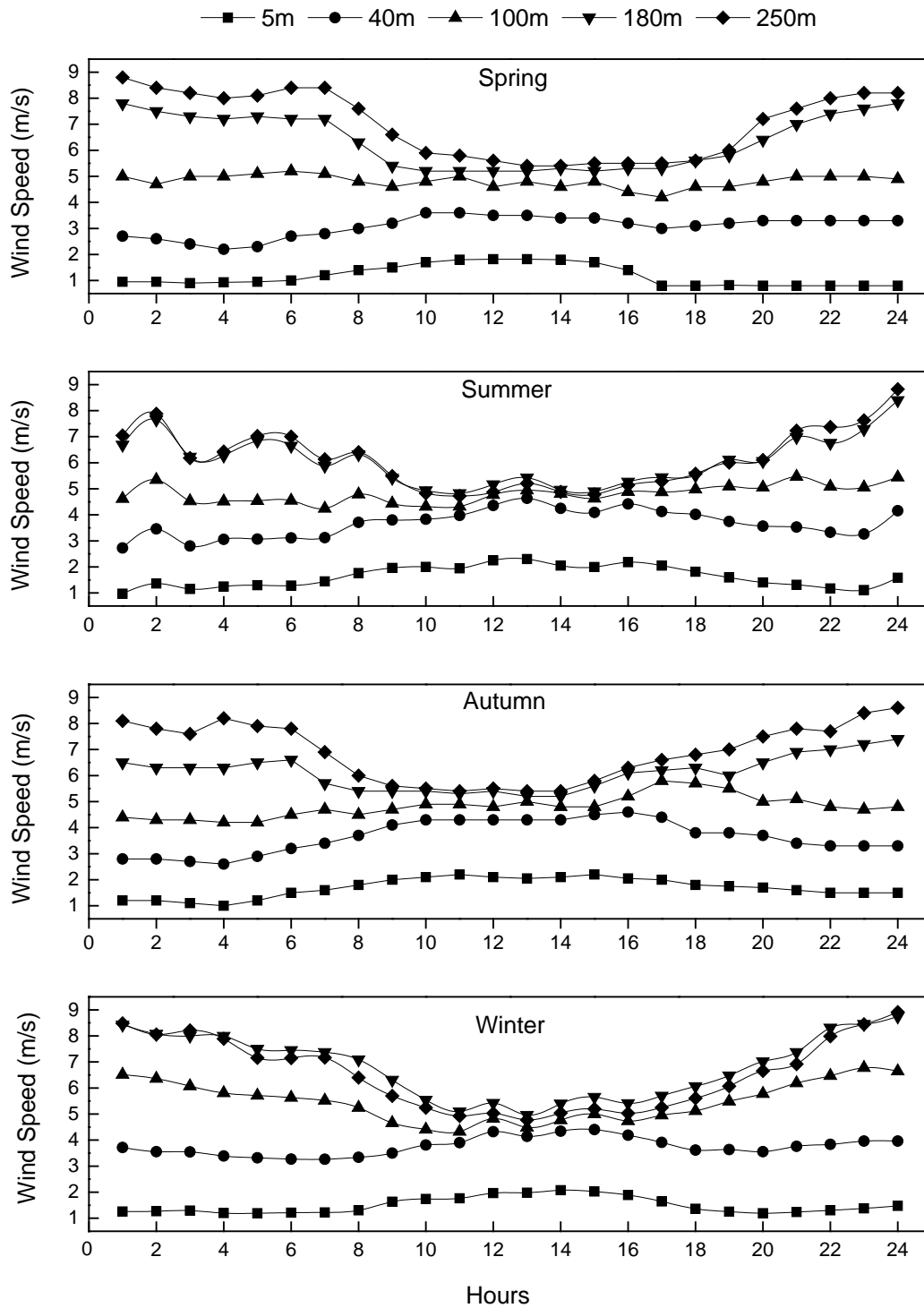
16 Surface layer has a remarkable effect on the diffusion of air pollutants. This layer is
17 strongly affected by the human behavior on the ground. Figure 1 presents on diurnal
18 variation of averaged wind speed in four seasons at different heights in Tianjin. The
19 four seasons were designated as March to May for spring, June-August for summer,
20 September-November for autumn, and December-February the next year for winter.
21 Diurnal variation patterns of wind speed were similar in each season. The wind speed
22 is high in daytime and low at night below 100m, whereas low wind speed in daytime
23 and high at night above 100m.

24 Figure 2 shows the vertical profile of wind speed and temperature in low atmosphere
25 under different stability. The gradient Richardson number(R_f) was used for
26 classifying the atmospheric stability conditions:

$$1 \quad R_i = \frac{g}{T} \left[\frac{\Delta T}{\sqrt{z_1 z_2} \ln \frac{z_2}{z_1}} + r_d \right] \times \left[\frac{\sqrt{z_1 z_2} \ln \frac{z_2}{z_1}}{\Delta u} \right] \quad (1)$$

2 Where, $\Delta T = T_2 - T_1$, $\Delta u = u_2 - u_1$, T_2 and T_1 are the measured temperatures at the
3 height of z_2 and z_1 , \bar{T} is the averaged temperature in the layer between level z_2 and
4 z_1 , u_2 and u_1 are the measured wind speed at levels z_2 and z_1 , g is the
5 gravitational acceleration, r_d is dry adiabatic lapse rate. According to the values of
6 R_i , three different conditions can be distinguished: $R_i \geq 0.1$ for stable condition,
7 $-0.1 < R_i < 0.1$ for neutral condition, and $R_i \leq -0.1$ for unstable condition.

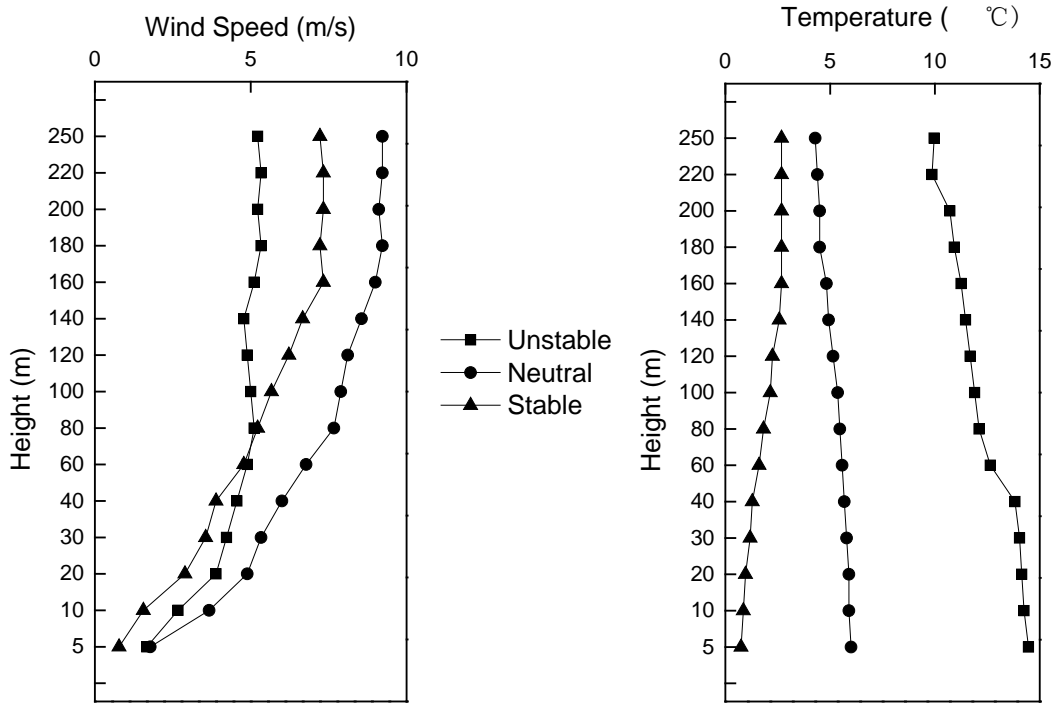
8 The atmospheric layer at 100-150m is considered as a transition layer, the variation
9 patterns of temperature and wind speed with height were different compared with the
10 upper and lower layers. Weak vertical gradient in the temperature profile was
11 observed over 100m. Similarly, small vertical gradient in wind speed was found over
12 150m.



1

2 Figure 1. Diurnal variation of averaged wind speed in each season at different heights

3



1

2 Figure 2. Vertical distribution profile of average wind speed and temperature in low
 3 atmosphere under different stability

4

5 3.2 The height of nocturnal planetary boundary and vertical variation of 6 turbulent intensity

7 The height of the planetary boundary layer (PBL), indicating the range of pollutants
 8 diffused by thermal turbulence in the vertical direction (Kim et al., 2007; Lena and
 9 Desiato, 1999), can be calculated by wind and temperature profiles (Seibert et al.,
 10 2000; Han et al., 2009). Based on the temperature profile observed at the tower, the
 11 vertical gradient of temperature was calculated as:

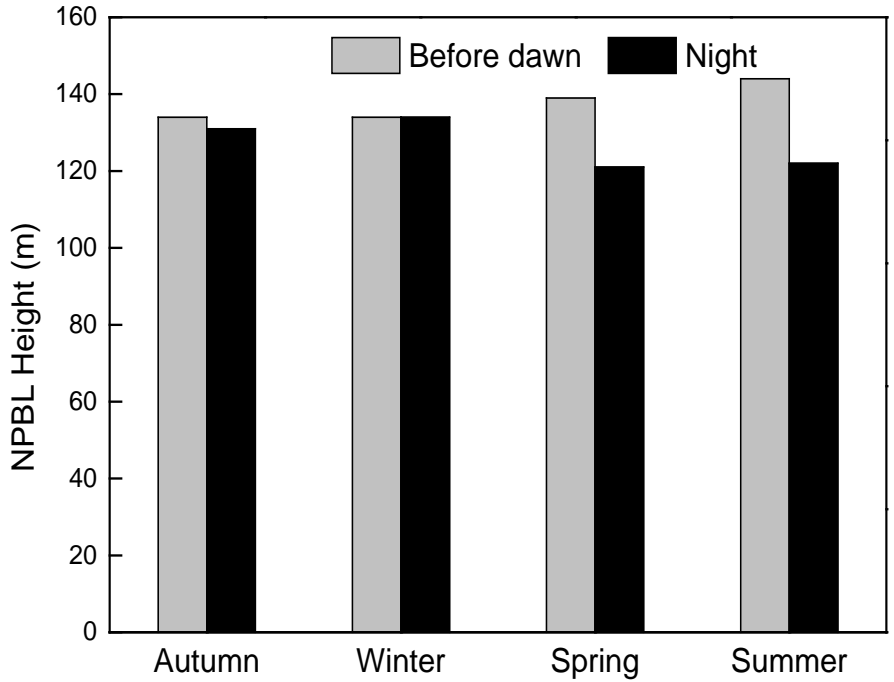
$$12 \quad \frac{\Delta T}{\Delta Z} = \frac{T(z+1) - T(z)}{Z(z+1) - Z(z)}$$

13 (2)

14 where $T(z+1)$ and $T(z)$ represent the measured temperatures at levels $z+1$ and
 15 z , $Z(z+1)$ and $Z(z)$ represent the altitudes at levels $z+1$ and z . The height

1 of the nocturnal planetary boundary layer (NPBL) is determined by the bottom of the
2 inversion, i.e. the layer in which temperature profile presents positive gradient. As
3 shown in Figure 3, the seasonal variation of the NPBL height is generally small, with
4 seasonal averaged NPBL height ranging from 114 to 142 m.

5



6

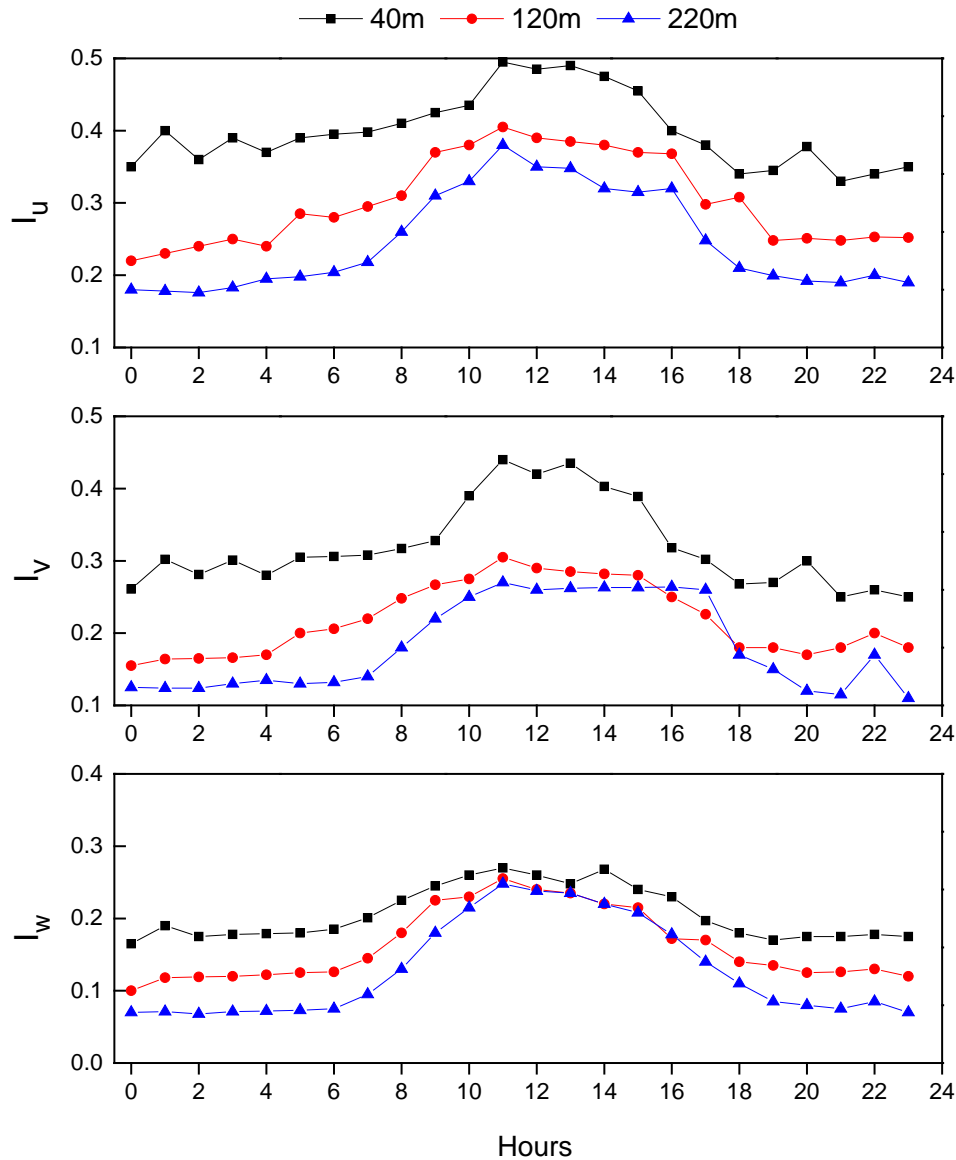
7 Figure 3. Averaged NPBL height in each season (before dawn 1:00-7:00; at
8 night:19:00-24:00)

9

10 In this study, hourly averaged $PM_{2.5}$ concentration measurement and twenty-four hour
11 PM_{10} filter sampling were conducted at four platforms. The heights of the 1st and 2nd
12 platform are inside the NPBL, the 3rd platform is located at the top of the NPBL, and
13 the 4th platform is generally outside the NPBL. Due to the dynamical stability of the
14 NPBL, air pollutants in surface layer are normally trapped inside the NPBL and rarely
15 mix with the pollutants outside the NPBL. Very different distribution characterizations
16 of PM were measured inside and outside the NPBL (See section 4).

1 Based on the observation data from the three dimensional ultrasonic anemometers, the
2 turbulent intensity were calculated. As a whole, the averaged diurnal variations of
3 turbulent intensity in each season(Supplemental Fig. S1) were reflecting the same
4 trends. The diurnal peaks appeared later and turbulent intensity was slightly weaker in
5 winter than in other seasons. Averaged diurnal variation of turbulent intensity at
6 different heights during the year of 2009 is shown in Fig. 4. Three dimensional
7 components of turbulent intensity decreased with increase in height. From the height
8 of 40m to 120 m, the u, v and w components of turbulent intensity reduced by 27%,
9 32% and 21%, respectively. From 120 to 220 m, the u, v and w components reduced
10 by 12%, 13% and 15%, respectively. The descending trend is more obvious from 40
11 to 120 m than that of from 120 to 220 m. This indicates that there were fully vertical
12 and horizontal turbulence exchanges below 120m of the tower, but relatively weaker
13 exchanges over 120m.

14



1

2 Figure 4. Averaged diurnal variation of three dimensional components of turbulent
 3 intensity at different heights (longitudinal turbulent intensity I_u , lateral turbulent
 4 intensity I_v , vertical turbulent intensity I_w)

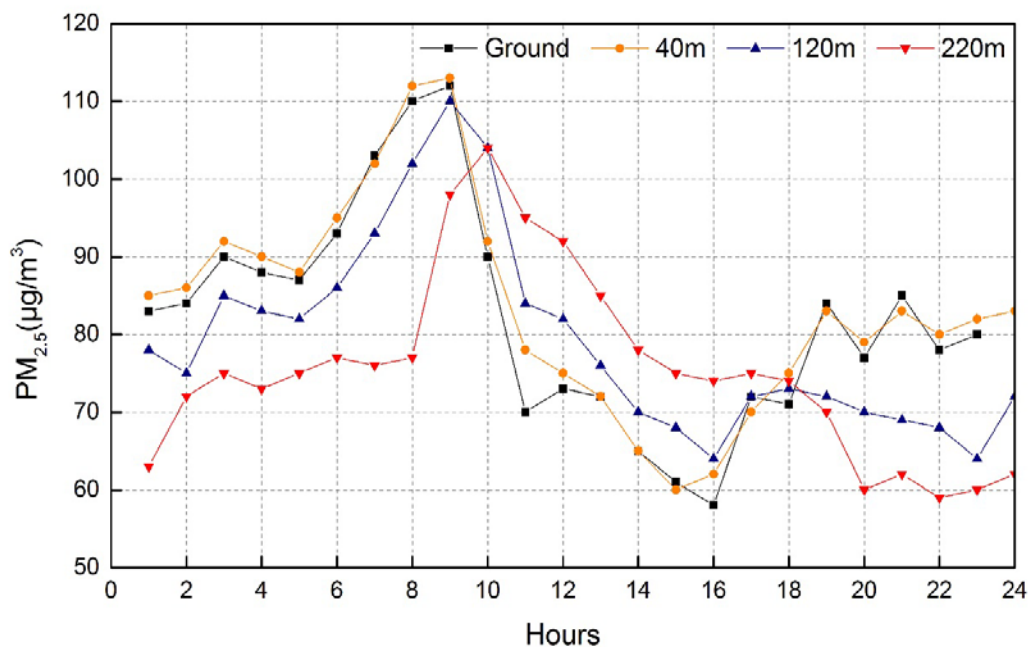
5

6 **4 Vertical distribution of $PM_{2.5}$ mass concentration**

7 The diurnal variation of $PM_{2.5}$ mass concentrations during the period from July 1 to
 8 September 30, 2009 is shown in Fig. 5. The vertical variation patterns of $PM_{2.5}$
 9 concentrations were quite different during the daytime and night resulting from a
 10 combination of diurnal variations of emissions and planetary boundary layer (PBL).
 11 After sunrise, the PBL height starts to rapidly increase, pollutants near the ground

1 gradually diffuse upward and the $PM_{2.5}$ concentration near the surface gradually
 2 decreases. At noontime, the mixing layer is fully developed with the averaged PBL
 3 height being about 1000-1200m. Among these 4 platforms (2 m, 40 m, 120 m and 220
 4 m), $PM_{2.5}$ concentration at 220m is the highest during noon-afternoon-time. In
 5 contrast, after 6 PM, the PBL height starts to rapidly decrease. The nocturnal
 6 planetary boundary layer(NPBL) height generally ranges from 100 m to 150 m(Fig.
 7 3). At the 1st and 2nd platform (2, 40 m), the measured PM are normally at inside of
 8 the NPBL. By contrast, the measurement platform at 220 m is generally outside the
 9 NPBL. The level 3 (120 m) is considered as at the transition zone between inside and
 10 outside of the NPBL. Due to the dynamical stability of the NPBL, the vertical mixing
 11 of pollutants between inside and outside of the NPBL is very weak. The surface
 12 emitted PM are normally trapped inside the NPBL, leading to the difference in the
 13 amount of aerosols below and above the NPBL. Among these 4 platforms, $PM_{2.5}$
 14 concentration at 220m during the night is the lowest. This indicates that the
 15 observation value of 220 m at night is less affected by local sources of emission and is
 16 largely attributed to regional scale pollution.

17



18

19 Figure 5. Vertical diurnal variation of $PM_{2.5}$ mass concentrations during the period

20 from July 1 to September 30, 2009

5 Vertical distributions of PM₁₀ concentration, composition and source apportionment

5.1 Vertical characteristics of PM₁₀ concentration

As mentioned in section 2.2, PM₁₀ filter samples were collected at the heights of 10, 40, 120 and 220 m. The daily concentrations at each sampling height were $139 \pm 45 \mu\text{gm}^{-3}$, $121 \pm 43 \mu\text{gm}^{-3}$, $110 \pm 39 \mu\text{gm}^{-3}$ and $79 \pm 37 \mu\text{gm}^{-3}$, respectively. These concentrations exhibited a general decreasing trend with the increase of height.

The height-to-height correlation coefficients of the variation of PM₁₀ concentration were calculated and listed in Table 1. All the pairwise correlation coefficients among 10, 40 and 120 m were higher than 0.9. However, the correlation coefficients between 220 m and other heights were obviously low. These results suggest that the influences of local emissions and local meteorological diffusion conditions on PM₁₀ concentrations are weaker at 220 m than that at lower levels.

Table 1. Height-to-height correlation coefficient of PM₁₀ concentration

	10 m	40 m	120 m	220 m
10 m	1.0	.		
40 m	0.96	1.0		
120 m	0.91	0.94	1.0	
220 m	0.72	0.76	0.85	1.0

5.2 Vertical characteristics of PM₁₀ chemical composition

Coefficients of divergence (CD) analysis (Wongphatarakul et al., 1998; Krudysz et al., 2009) was used in this study to assess vertical variability of chemical elements in PM₁₀ filter samples collected at 4 heights. The CD values provide information on the degree of uniformity between sampling sites and is defined as

$$CD_{jk} = \sqrt{\frac{1}{\rho} \sum_{i=1}^{\rho} \frac{(x_{ji} - x_{ik})^2}{x_{ji} + x_{ik}}} \quad (3)$$

where, x_{ji} is the average concentration of the i th element at j th height. j and k represent the two sampling heights, and ρ is the number of elements. When the species concentrations at two sampling sites were similar to each other, the CD values would approach 0. On the other hand, as the two species concentrations diverge the CD value will approach 1 (Hwang et al., 2008).

The pair-wise CD values for four heights are shown in Table 2. The pair-wise CD values among 10, 40 and 120m are lower than 0.2, illustrating that the element profiles of these three heights were similar to each other. While, the CD values between 220m and the other three levels were obviously high. This may be resulted from that chemical elements in the PM₁₀ filter samples collected at 220m were mainly originated from regional-scale sources.

Table 2. Pair-wise CD values at different heights

	10 m	40 m	120 m
40 m	0.10		
120 m	0.15	0.11	
220 m	0.33	0.30	0.59

14

The concentration of chemical composition in ambient PM₁₀ filter samples collected at 4 heights are shown in Table 3. Al, Si, Ca, OC, EC, Cl⁻, NO₃⁻ and SO₄²⁻ have higher concentration levels than other species. Al can be used as a source marker of coal combustion (Hopke, 1985) ; Al and Si are the markers of soil dust (Liu et al., 2003), Ca is mainly emitted from cement dust (Shi et al., 2009) ; EC can be identified as vehicle exhaust emission (Li et al., 2004) ; Cl⁻ is the marker for sea salt (Li et al., 2004); and NO₃⁻ and SO₄²⁻ are the markers of secondary nitrate and sulfate (Liu et al.,

2003). Higher concentrations were found at lower sampling heights for almost all species (NO_3^- had the highest value at 120 m). Unlike the species concentration, the vertical distribution of species percentages (%) shows different patterns. Similar fraction levels were observed at the four heights for Al and Si. For Ca and EC, higher values were observed at lower sampling sites. The percentages of OC at 220 m were obviously higher than those at 120 m. This might imply that the influence of local sources on OC was weaker and the contributions from secondary and regional sources were larger at 220 m. The OC/EC ratios increased gradually from 10 m to 220 m. This might be due to a relatively higher percentage of SOC in OC at higher heights as results of the formation and regional transport of SOC (Strader et al., 1999). Similarly, the higher sampling sites obtained higher fractions (%) for NO_3^- and SO_4^{2-} (the highest percentage of NO_3^- were observed at 120m). These trends suggest that the impact of primary sources from the ground decreased with the increase of height, while the impact of secondary sources mainly influenced by regional sources becomes more prominent.

Table 3. The concentration of chemical composition in ambient PM_{10} at 4 height sampling sites ($\mu\text{g m}^{-3}$)

	10m		40m		120m		220m	
	mean	sd ^a	mean	sd	mean	sd	mean	sd
Na	1.60	0.71	1.34	0.58	1.28	0.48	0.89	0.41
Mg	1.51	0.54	1.29	0.92	0.99	0.52	0.54	0.36
Al	6.3	2.5	5.9	2.1	4.9	1.7	4.0	1.7
Si	8.5	4.6	6.8	2.9	6.4	2.8	4.9	2.8
P	ND	ND	ND	ND	ND	ND	ND	ND
K	1.41	0.72	1.02	0.44	1.11	0.68	0.70	0.35
Ca	7.1	2.8	5.1	2.0	4.6	2.2	2.5	1.6
Ti	0.23	0.12	0.19	0.12	0.24	0.20	0.29	0.53
V	ND	ND	ND	ND	ND	ND	ND	ND
Cr	0.04	0.03	0.04	0.03	0.05	0.04	0.04	0.04
Mn	0.09	0.05	0.06	0.03	0.06	0.03	0.04	0.02
Fe	2.51	1.22	2.08	1.21	1.92	1.09	1.09	0.80
Ni	0.01	0.02	0.01	0.01	0.02	0.03	0.03	0.05
Co	0.01	ND	ND	ND	ND	ND	0.01	0.01
Cu	0.20	0.17	0.14	0.22	0.09	0.13	0.02	0.03
Zn	0.69	0.32	0.60	0.31	0.55	0.28	0.27	0.16

Br	ND	ND	ND	ND	ND	ND	ND	ND
Ba	ND	ND	ND	ND	ND	ND	ND	ND
Pb	0.06	0.06	0.06	0.06	0.05	0.05	0.03	0.03
OC ^a	13.5	6.2	10.8	4.6	9.6	3.8	7.3	3.1
EC ^a	7.0	2.2	5.3	2.0	4.4	1.8	3.0	1.6
NH ₄ ⁺	6.2	3.5	6.3	3.4	6.9	3.1	5.7	4.0
Cl ⁻	6.4	5.3	5.6	4.1	5.0	3.0	1.7	1.2
NO ₃ ⁻	18.0	12.5	16.9	10.9	18.9	10.1	13.3	11.4
SO ₄ ²⁻	27.4	20.6	26.1	17.5	25.3	16.4	19.7	16.2
OC/EC	1.91	2.79	2.03	2.26	2.20	2.10	2.40	1.90
PM ₁₀	140	48	120	44	108	41	80	39

1 ^a sd: standard deviation; OC: organic carbon; EC: element carbon.

2

3 **5.3 Vertical characteristics of PM₁₀ sources**

4 In order to understand the vertical characteristics of PM₁₀ sources, the chemical mass
5 balance (CMB) model was applied for source apportionment at all four sampling
6 heights. The CMB model, a useful receptor model, has been extensively used to
7 estimate source categories and contributions to the receptor based on the balance
8 between sources and the receptor (Chow et al., 2007; Watson et al., 2008). Further
9 details of CMB can be found in the relative literature (Watson et al., 1984; Watson et
10 al., 2002; USEPA, 2004). The dataset of chemical composition in the PM₁₀ samples
11 during the measurement period and the source profiles reported in our previous
12 works(Bi, et al., 2007) were used in the CMB modeling.

13 Six source categories (coal combustion, crustal dust, cement dust, vehicle exhaust,
14 secondary sulfate and secondary nitrate) and their source contributions ($\mu\text{g m}^{-3}$) and
15 percentage contributions (%) estimated by the CMB model are listed in Table 4. The
16 estimated source contributions ($\mu\text{g m}^{-3}$) of all the sources showed a downward trend
17 with the increase of height. Whereas the percentage contributions (%) of secondary
18 sources (secondary sulfate and nitrate) presented a generally increasing trend with the
19 increase in height. This might be due to the fact that for the secondary sources the
20 particulate sizes are relatively smaller and the residence time of fine particle is longer.

1 Generally, secondary sources can obtain stronger influence from regional
 2 contributions (Gu et al., 2011). That is to say, PM at higher heights obtain more
 3 regional contributions. And, to some extent, this could reflect the characteristics at the
 4 regional scale.

5

6 Table 4. Source contributions and percentage contributions at four different heights

		coal combustion	crustal dust	cement dust	vehicle exhaust	secondary sulfate	secondary nitrate	TOT
contribution ($\mu\text{g m}^{-3}$)	10m	17	16	14	20	34	23	140
	40m	16	13	10	17	33	21	120
	120m	14	12	8	15	32	24	108
	220m	12	9	4	12	25	17	80
percentage (%)	10m	12	11	10	14	24	16	88
	40m	13	11	8	14	27	18	90
	120m	13	11	8	14	29	22	97
	220m	14	11	5	15	31	21	97

7

8 **6 Vertical variation of periodicity for the time series of PM_{2.5}**
 9 **concentrations**

10 The periodic characteristics of particulate concentration and meteorological variables
 11 can reflect different scales of atmospheric processes. In this paper, the vertical
 12 variation period of PM_{2.5} mass concentrations were analyzed.

13 Time series of atmospheric pollutant concentration could be decomposed into baseline
 14 and short-term components. Using the filtering method, short-term fluctuations
 15 associated with the influence of local-scale pollution and dispersion conditions can be
 16 extracted from the original measurements. After the removal of local-scale effects, the
 17 time series of pollutant concentrations can be reconstructed to reflect the regional
 18 scale influence.

19 **6.1 Filtering method**

20 The wavelet transform can be used to analyze time series that contain nonstationary
 21 signals at many different frequencies. In this paper, we chose the Morlet wavelet
 22 which is extensively used in studies of climate change and turbulence power spectrum

1 analysis (Torrence and Compo, 1998). The normalization mother wavelet is shown in
 2 Eq. (4).

$$3 \quad \psi_0(\eta) = \pi^{-1/4} e^{i\omega_0\eta} e^{-\eta^2/2} \quad (4)$$

4 where η is the nondimensional time parameter and ω_0 is the nondimensional
 5 frequency. The wavelet filter time series over a set of scales can be calculated by:

$$6 \quad x_n = \frac{\delta j \delta t^{1/2}}{C_\delta \psi_0(0)} \sum_{j=0}^J \frac{R\{W_n(s_j)\}}{s_j^{1/2}} \quad (5)$$

7 where δj is the spacing between the discrete scales, and δt is the sampling interval.

8 S_j is a set of scales related to the frequency ω . C_δ and $\psi_0(0)$ are both constants.

$$9 \quad \omega = \frac{\omega_0 + \sqrt{2 + \omega_0^2}}{4\pi s} \quad (6)$$

10 The reconstruction then gives:

$$11 \quad C_\delta = \frac{\delta j \delta t^{1/2}}{\psi_0(0)} \sum_{j=0}^J \frac{R\{W_\delta(s_j)\}}{s_j^{1/2}} \quad (7)$$

12 According to the conservation of total energy under the wavelet transform and the
 13 equivalent of Parseval's theorem for wavelet analysis, the variance of the time series
 14 is:

$$15 \quad \sigma^2 = \frac{\delta j \delta t}{C_\delta N} \sum_{n=0}^{N-1} \sum_{j=0}^J \frac{|W_n(s_j)|^2}{s_j} \quad (8)$$

16 Both Eqs. (7) and (8) should be used to check wavelet routines for accuracy and to
 17 ensure that sufficiently small values of s_0 and δj have been chosen. The values of
 18 the above parameters are given in Table 5.

19 As discussed above, the wavelet transform is essentially a bandpass filter. By
 20 summing over a subset of the scales in Eq. (5), a wavelet-filtered time series can be
 21 constructed as follows:

$$x_n' = \frac{\delta j \delta t^{1/2}}{C_\delta \psi_0(0)} \sum_{j=j_1}^{j_2} \frac{R\{W_n(s_j)\}}{s_j^{1/2}} \quad (9)$$

This filter has a response function given by the sum of the wavelet functions between scale j_1 and j_2 .

Table 5. Values of the parameters of the Morlet transform in this study

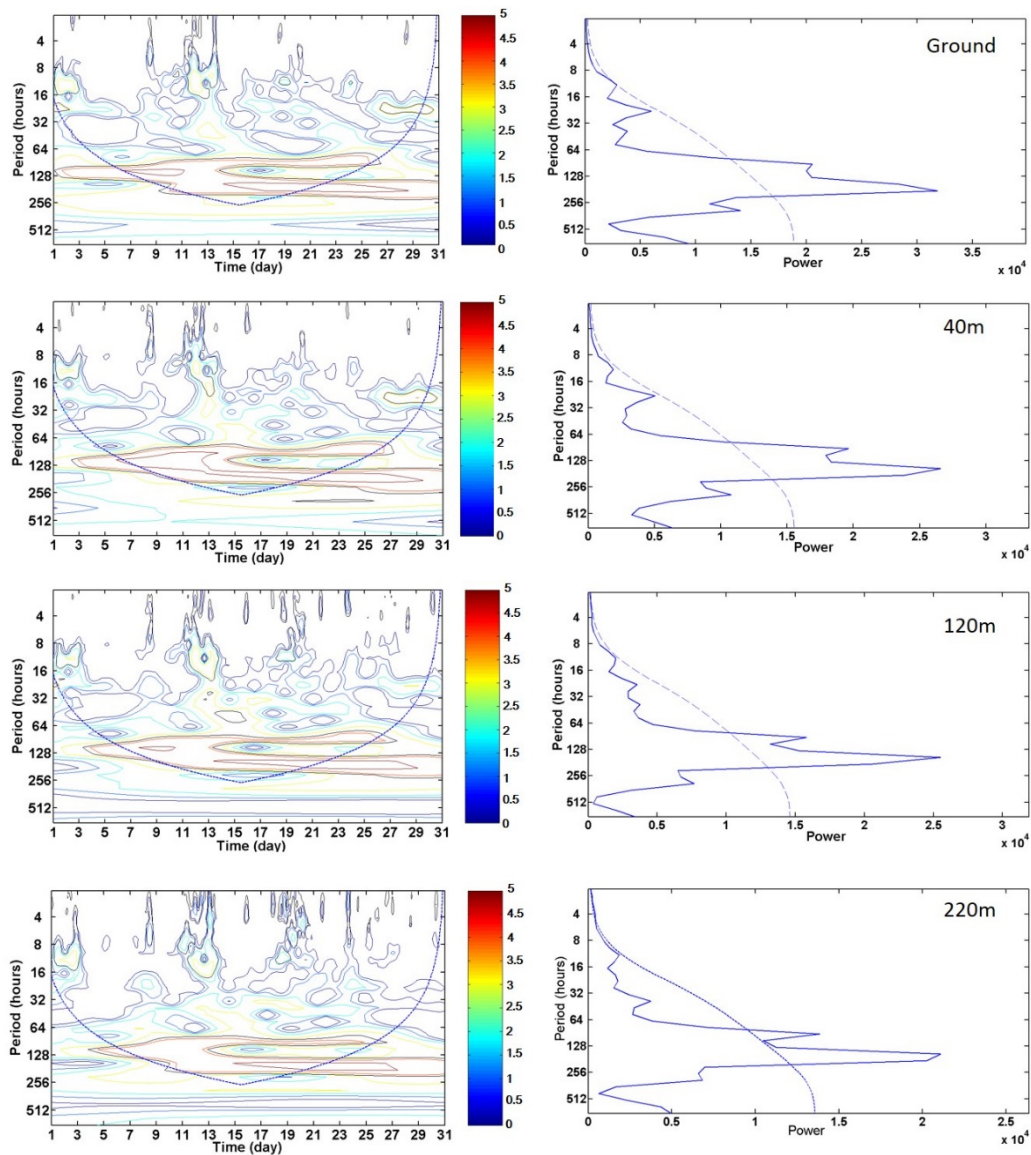
C_δ	ψ_0	s_0	δt	δj	ω_0
0.776	$\pi^{-1/4}$	$2\delta t$	2	0.25	6.0

6.2 Fluctuation spectrum analysis of PM_{2.5} concentration time series at different heights

The fluctuation spectrum distribution of hourly mass concentrations of PM_{2.5} on the ground and at the height of 2, 40, 120 and 220 m were analyzed in this paper. The missing data in the time series was computed by interpolation. Because of low proportions and unconcentrated distributions in the missing data, little human interference was brought to the spectral composition of the original time series. For better comparison, normalization (standard variance 1, mean 0) of the original time series was necessary prior to power spectrum analysis.

The local and global wavelet power spectrum contours for the time series of PM_{2.5} concentrations at different heights in August are shown in Fig. 6. Contours are expressed as $\log_2(|W_n(s)|^2)$ because of large magnitudes. Area inside the thick black solid line passes the red noise standard spectral test with the 5% significance level. Area outside the blue dotted line was excluded from analysis because of poor reliability from the cone of influence, where edge effects become important. The global wavelet spectrum $\overline{W^2}(s)$, which reflects characteristics of the pollutant concentration time series in the frequency domain, was obtained by calculating the average of local wavelet spectrums $|W_n(s)|^2$ over the entire sampling time domain. The solid line is the global wave spectrum for the corresponding time series. The dashed line is the 5% significance level, the upper area of which passes the red noise standard spectral test at the 5% significance level.

1 The global wavelet power spectrum of PM_{2.5} mass concentration shows that
 2 fluctuations of 6-10 days (related to weather process and regional-scale pollution) are
 3 significant at each observation height, while fluctuations of 12-24 hours (mainly
 4 concerned with the daily variation of atmospheric boundary layer and local pollution
 5 emissions by human activities) are significant only on ground level. For the
 6 fluctuations of PM_{2.5} mass concentration, wave energy of 6-10 days period reduces
 7 with the increase of height. In terms of the local power spectrum, 12-24 hours period
 8 can be observed in a few days on the ground. But with the increase of height, the
 9 power of 12-24 hours period became weaker, only 10%-30% of that on the ground.



10

11

12 Figure 6. Local (left figure) and global (right figure) wavelet power spectrum of

13 PM_{2.5} mass concentration at different heights in August, 2009

1

2 **7 Determination of regional background concentration of particulate** 3 **matter**

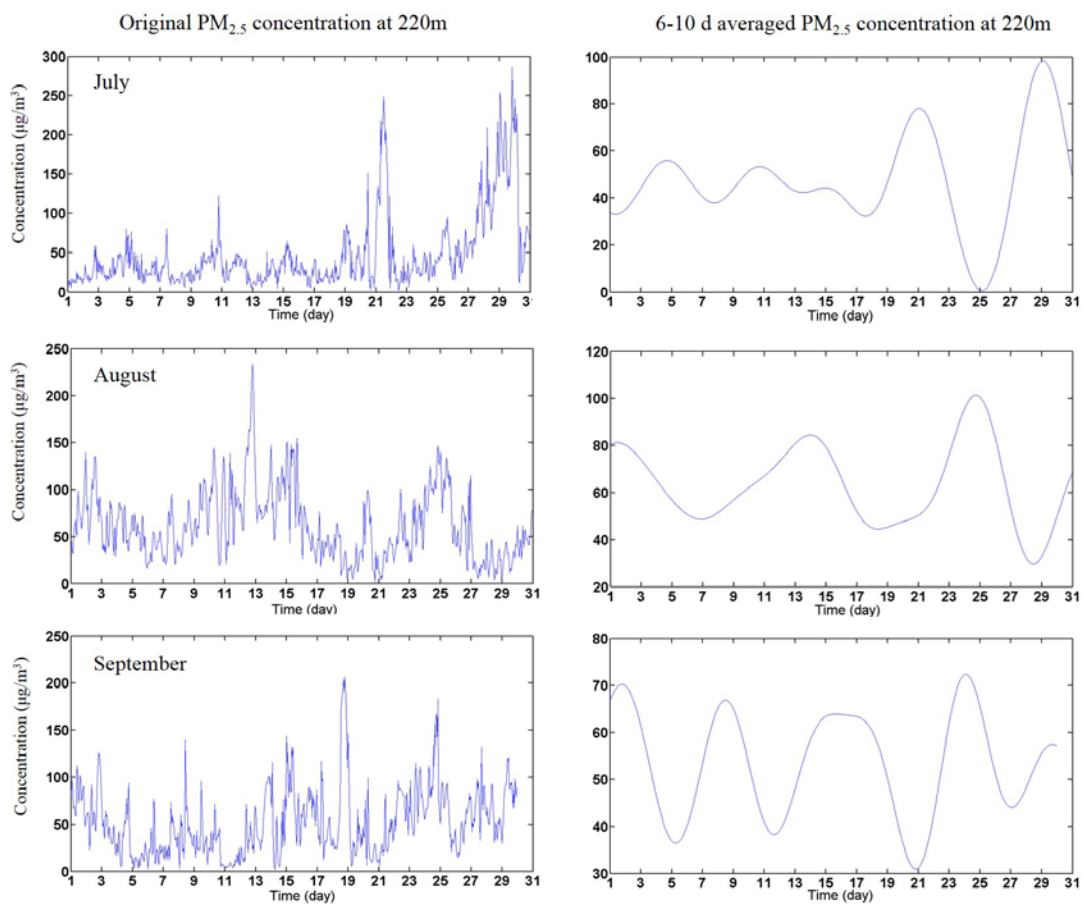
4 Regional PM background concentration can hardly be measured directly. Original PM
5 concentration time series measured on the ground reflect a combination of influence
6 from local pollution and regional-scale pollution. This study is expected to give a way
7 to characterize the regional pollution contribution and to evaluate regional
8 background PM concentration levels. According to the above research concerning the
9 vertical distribution characteristics of particle size, chemical composition and
10 pollution sources, the atmospheric boundary layer structure, as well as the fluctuation
11 power spectrum analysis of particle mass concentration, the measurement height
12 influenced relatively less by local pollution emission was determined and impacts
13 from local-scale pollution on the short-term fluctuations have been removed from the
14 original PM concentration by wavelet transformation. The nocturnal PM_{2.5} mass
15 concentration time series with the 6-10 days period at the observation height of 220 m
16 were extracted to characterize the regional background concentration, which mainly
17 associated with the regional scale pollution within 10² km away from the
18 measurement tower.

19 Time series of PM_{2.5} hourly concentration before and after the filtering was presented
20 in Fig. 7. Due to short-term fluctuations of pollution emission and local diffusion
21 conditions, observation errors, and etc., the original PM_{2.5} concentration time series
22 presents violent oscillation. Using wavelet transformation, the nocturnal PM_{2.5} mass
23 concentration time series with the 6-10 days period at the height of 220m was
24 extracted from the original time series. After the filtering, impacts from local-scale
25 pollution and diffusion conditions on the short-term fluctuations were considered to
26 be removed. Thus regional-scale pollution and synoptic-scale weather conditions were
27 better represented in the remaining part compared with the original PM concentration
28 time series.

29 The swings in the PM_{2.5} concentration data(shown in Fig. 7) were mainly resulted
30 from several meteorological processes during the measurement. According to the
31 meteorological dataset of the observation station(WMO Id.No. 54517,), precipitation
32 processes were recorded during the period of 22-24 July, with the amounts of rainfall

1 ranged from 3.2 to 94.6mm, followed by a rapid decrease in PM_{2.5} concentration on
 2 25 July due to consequent cleaning of the air. Then, beginning on 26 July, mist paired
 3 with calm winds caused a build-up of PM_{2.5} concentration until July 29. Similar
 4 meteorological processes were reported during the period of 22-25 of August, 4-9 and
 5 20-25 of September, which resulted in the cycle of cleaning and build-up of air
 6 pollutants.

7 According to the method proposed in this paper, in Tianjin, the averaged regional
 8 background PM_{2.5} concentrations in July, August and September, 2009 were 40 ± 20
 9 $\mu\text{g m}^{-3}$, $64 \pm 17\mu\text{g m}^{-3}$ and $53 \pm 11\mu\text{g m}^{-3}$, respectively.



10

11 Figure 7. Time series of PM_{2.5} hourly concentration before and after the filtering

12

13 **8 Summary and conclusions**

14 It is crucial for studying regional-scale PM pollution and for the development of
 15 efficient joint control policy to improve understanding of the regional background
 16 concentration of PM. The purpose of this study is to characterize the regional

1 pollution contribution and to evaluate regional background PM concentration levels.
2 However, regional background concentration can hardly be measured directly.
3 Original PM concentration time series measured on the ground reflect a combination
4 of influence from local pollution and regional-scale pollution. A method to estimate
5 regional background PM concentration is proposed in this paper, based on the vertical
6 variation periodic characteristics of particle mass concentration, the atmospheric
7 boundary layer structure, as well as the vertical distribution of chemical composition
8 and pollution source apportionment .

9 Based on a 255 m meteorological tower, the vertical thermodynamic and dynamic
10 characteristics of the atmospheric boundary layer in Tianjin was observed. The
11 atmospheric layer at 100-150m is considered as a transition layer, the variation
12 patterns of temperature and wind speed with height were different compared with the
13 upper and lower layers. Weak vertical gradient in the temperature profile was
14 observed over 100m. Similarly, small vertical gradient in wind speed was found over
15 150m.. The turbulent intensity decreased with increase in height and the descending
16 trend is more obvious from 40 to 120 m than that of from 120 to 220m, which
17 indicates that there were fully vertical and horizontal turbulence exchanges below
18 120m of the tower, but relatively weaker exchanges over 120m. Seasonal averaged
19 nocturnal planetary boundary layer height ranges from 114 to 142 m. The observation
20 height of 220 m is just outside the NPBL, which indicates that the observation value
21 of PM concentration at 220 m at night is less affected by local primary sources near
22 the ground and is largely contributed by regional scale pollution.

23 The vertical distribution of chemical composition in PM₁₀ filter samples also suggests
24 that the impact of primary sources near the ground decreased with height, whereas the
25 impact of secondary sources mainly influenced by regional sources became more
26 prominent. The vertical distribution of percentage was different for various species.
27 Similar percentage levels were observed at the four different heights for Al and Si.
28 For the Ca and EC fractions, higher values were observed at lower sampling sites. The
29 percentages of NO₃⁻, SO₄²⁻ and OC, and the OC/EC ratios were obviously higher at
30 higher sites. Source apportionment for ambient PM₁₀ showed that the percentage
31 contributions of secondary sources obviously increased with height, while the
32 contribution of cement dust decreased with height. PM at higher height obtained more
33 regional contributions, and to some extent, it could reflect the characteristics of the

1 regional scale.

2 The periodic characteristics of $PM_{2.5}$ mass concentration can reflect different scales of
3 atmospheric processes. In terms of global wavelet power spectrum of $PM_{2.5}$ mass
4 concentration, fluctuations of 6-10 days, related to weather processes and
5 regional-scale pollution, were significant at each observation height. While
6 fluctuations with 12-24 hours period, mainly concerned with the daily variation of
7 atmospheric boundary layer and local pollution emissions by human activities in the
8 surface layer, were significant only on ground level. In terms of the local power
9 spectrum, 12-24 hours period can be observed in a few days on the ground. But with
10 the increase of height, the power of 12-24 hours period became weaker, only 10-30%
11 of that on the ground.

12 According to the above research, the nocturnal $PM_{2.5}$ mass concentration time series
13 with the 6-10 days period at the measurement height of 220m can be regarded as
14 regional background concentration, which mainly associated with the regional scale
15 pollution within 10^2 km away from the measurement tower. Using wavelet
16 transformation and filtering, the nocturnal $PM_{2.5}$ mass concentration time series with
17 the 6-10 days period at the height of 220m was extracted from the original time series.
18 After removing the impacts from local-scale pollution and diffusion conditions on the
19 short-term fluctuations, regional-scale pollution and synoptic-scale weather conditions
20 were better represented in the remaining part compared with the original PM
21 concentration time series. According to the method proposed in this paper, in Tianjin,
22 the averaged regional background $PM_{2.5}$ concentrations in July, August and September,
23 2009 were $40 \pm 20 \mu g m^{-3}$, $64 \pm 17 \mu g m^{-3}$ and $53 \pm 11 \mu g m^{-3}$, respectively.

24 We attempted to put forward a new method to estimate the regional background
25 concentration of PM. Background PM concentrations are not constant but varying
26 with space and time. In future research, more analysis on the characteristics of the
27 urban boundary layer, vertical distribution of PM composition and source
28 apportionment in different seasons and meteorological conditions will be done, and
29 background concentration ranges of $PM_{2.5}$ for given time periods and meteorological
30 conditions will be obtained.

31

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7

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