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Humidity and PM_{2.5} composition determine atmospheric light extinction in the arid region of northwest China

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ABSTRACT

Atmospheric visibility can directly reflect the air quality. In this study, we measured water-soluble ions (WSIs), organic and element carbon (OC and EC) in PM_{2.5} from September 2017 to August 2018 in Urumqi, NW China. The results show that SO₄²⁻, NO₃⁻ and NH₄⁺ were the major WSIs, together accounting for 7.32%–84.12% of PM_{2.5} mass. Total carbon (TC=OC+EC) accounted for 12.12% of PM_{2.5} mass on average. And OC/EC > 2 indicated the formation of secondary organic carbon (SOC). The levels of SO₄²⁻, NO₃⁻ and NH₄⁺ in low visibility days were much higher than those in high visibility days. Relative humidity (RH) played a key role in affecting visibility. The extinction coefficient (b_{ext}) that estimated via Koschmieder formula with visibility was the highest in winter ($1441.05 \pm 739.95 \text{ Mm}^{-1}$), and the lowest in summer ($128.58 \pm 58.00 \text{ Mm}^{-1}$). The b_{ext} that estimated via IMPROVE formula with PM_{2.5} chemical component was mainly contributed by (NH₄)₂SO₄ and NH₄NO₃. The b_{ext} values calculated by both approaches presented a good correlation with each other ($R^2 = 0.87$). Multiple linear regression (MLR) method was further employed to reconstruct the empirical regression model of visibility as a function of PM_{2.5} chemical components, NO₂ and RH. The results of source apportionment by Positive Matrix Factorization (PMF) model showed that residential coal combustion and vehicle emissions were the major sources of b_{ext} .

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Introduction

Air pollution in China has become serious owing to the rapid development of economy, industrialization and urbanization.

A series of PM_{2.5} pollution events arouse worldwide attention. The chemical composition of PM_{2.5} is extremely complex and mainly composed of water-soluble ions, organic matter, black carbon, and heavy metals. PM_{2.5} not only has adverse effect on human health (Palomo, 1999; Raizenne et al., 1996; Pope et al., 2009) but also directly or indirectly affects atmospheric chemistry, visibility, acid deposition, cloud and pre-

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cipitation formation as well as radiation balance (Zhang et al., 2012; Kaufman et al., 2002; Liu et al., 1999).

In recent decades, haze events frequently happened in China's urban areas, with PM_{2.5} concentrations largely exceeding the national standard (Tan et al., 2016; Zhang et al., 2011, 2017; Jun et al., 2015). Low visibility is associated with high concentrations of suspended particles, especially fine particles (< PM_{1.0}). The absorbing aerosol has a large Ångström exponent (AAE) value, 1.58 at 470/660 nm, much larger than the scattering aerosols (SAE) (Zhuang et al., 2017). The chemical composition, size and concentration of particle matter in the atmosphere affect visibility. Zou et al. (2018) found that ~30% of the light extinction coefficient came from secondary aerosols. Many previous studies have found that SO₄²⁻, NO₃⁻ and black carbon in PM_{2.5} aerosol have strong scattering and absorbing effects (Lowenthal and Kumar, 2012; Hand et al., 2002; Kim et al., 2001; Chen and Bond, 2009; Pathak et al., 2009; Wu et al., 2017a). In addition, other species in aerosols also have an impact on visibility. Wang et al. (2018) studied the atmospheric particle of the Tibetan plateau and found that rBC (refractory Black Carbon) had strong light absorption properties, and the rBC coated on particles was positively correlated with the light absorption intensity. The total light absorption contribution of brown carbon to aerosol could reach 20–40% at the wavelength of 350 nm, indicating that the light absorption contribution of brown carbon in the short-wave region cannot be ignored (Chen and Bond, 2009). However, BC is not the only particle that absorbs sunlight, humus, humus-like substances and bio-aerosols have also been verified as sunlight absorbing particles (Andreae and Gelencsér, 2006). The hygroscopicity of water-soluble substances in PM_{2.5} (mainly including water-soluble ions, water-soluble organic carbon and inorganic carbon) has a negative impact on visibility.

At present, the research on the optical characteristics and chemical components of particle in China mainly focuses on the economically developed regions such as Beijing-Tianjin-Hebei, Yangtze River and Pearl River Deltas (Tao et al., 2015; Han et al., 2015a, 2015b; Cui et al., 2016; Xia et al., 2017), while there have been few studies focusing on visibility in western China. Located in the Eurasian continental bridge economic belt, Urumqi is an important economic center in western China. With the sustained and rapid economic development, air pollution has become more and more severe. Moreover, Urumqi is surrounded by mountains and has arid climate (Turap et al., 2019), resulting in secondary inorganic ions and carbonaceous aerosols contribute largely to fine particle. The chemical species in PM_{2.5} can affect atmospheric visibility. Therefore, it is crucial to study the relationship between the chemical composition of PM_{2.5} and atmospheric visibility.

This study characterized daily PM_{2.5} samples collected in the southern Urumqi during Sep.2017 to Aug.2018. And light extinction coefficients was estimated by IMPROVE formula. Multiple linear regression (MLR) was used to elucidate the correlation of visibility with the chemical components of PM_{2.5}. The sources of atmospheric extinction were also attributed with the Positive Matrix Factorization (PMF) model.

1. Experimental method

1.1. Sample collection

PM_{2.5} samples were collected on the roof of No.9 building (43° 46' 29"N, 87° 37' 18"E) of Xinjiang University in Urumqi (Fig. 1). The sampling site was about 25 m high and is surrounded by residential, traffic and construction emissions and representative of urban Urumqi. PM_{2.5} samples were collected from Sep. 2017–Aug. 2018. Daily 22-hr sampling from 10:00 a.m. to

next day at 8:00 a.m. were carried out continuously for seven days every month. During each sampling episode, meteorological parameters, visibility and concentrations of gas contaminants were recorded (<https://www.aqistudy.cn/>). A blank sample was collected in each month. PM_{2.5} samples were collected on quartz fiber filter (203 mm × 254 mm, Whatman, British). All filters wrap in aluminum foil and prebaked at 450 °C for 4 hr to remove absorbed organic material. PM_{2.5} samples were collected using high-volume air sampler (TH-1000, Wuhan Tianhong instruments Co., Ltd, China) with a flow at 1.05 m³/min. After each sampling, the filter wrap in the aluminum foil again and stored under –20 °C to prevent evaporation of volatilized components.

1.2. Gravimetric weighing

All filters were weighted before and after sampling to obtain particle mass concentrations by using electronic microbalance (AB204-S, Mettler Toledo, Switzerland) after 24 hr equilibration at 20 ± 2 °C and relative humidity at 45%–55%. Each filter was weighed at least three times before and after sampling respectively. The difference between replicate was not exceeding 0.00003 g. If not so, the filters would be equilibration for 24 hr and weighted again.

1.3. Chemical analysis

A punch of each PM_{2.5} sample was placed into centrifuge tube and submerged in 10 mL of ultrapure water (18.2 MΩ/cm). Then an ultrasonic bath was operated for 20 min at low temperature (<10 °C), followed by centrifuge (11,000 r/min) for 10 min. The extract solutions were filtered with a polytetrafluoroethylene membrane (0.45 μm pore size), transferred into cleaned glass bottles, and then analyzed by Ion chromatography (883 Basic IC Plus, Metrohm, Switzerland) immediately. Five cations (Na⁺, NH₄⁺, K⁺, Mg²⁺, Ca²⁺) and three anions (SO₄²⁻, NO₃⁻, PO₄³⁻, Cl⁻, F⁻) were determined. In this study, the detection limits of Na⁺, NH₄⁺, K⁺, Mg²⁺, Ca²⁺, SO₄²⁻, NO₃⁻, PO₄³⁻, Cl⁻ and F⁻ are 0.014, 0.014, 0.046, 0.017, 0.041, 0.018, 0.032, 0.136, 0.022 and 0.026 μg/m³, respectively. The recovery of each ion was 90%–105%.

A punch of 0.5 cm² from each filter was analyzed for OC and EC using DRI Model 2015 Thermal/Optical Carbon Analyzer (Desert Research Institute, USA), using the IMPROVE_A protocol to measure OC and EC (Chow et al., 2001). The method detection limits (MDLs) for the OC and EC were 0.05 μg/m³, respectively. Replicate measurements were performed every ten samples, and the difference between replicate was < 10%.

1.4. IMPROVE equation

The IMPROVE formula reflects the relationship between the chemical composition of the aerosol and the atmospheric extinction (absorption and scattering). Assuming an externally mixed aerosol model consisting of large and small size modes, and the formula for calculating the extinction coefficient in urban atmosphere is as:

$$b_{\text{ext}} = b_{\text{ap}} + b_{\text{sp}} + b_{\text{ag}} + b_{\text{sg}} \quad (1)$$

where, b_{ap} , b_{sp} represent the absorption and scattering coefficients of aerosol, b_{ag} and b_{sg} absorption and scattering of gas, respectively.

$$b_{\text{ap}} = 10 \times [\text{EC}] \quad (2)$$

$$b_{\text{sp}}(\text{revised}) = 2.2 \times f_s(\text{RH}) \times [(\text{NH}_4)_2\text{SO}_4]_{\text{small}}$$

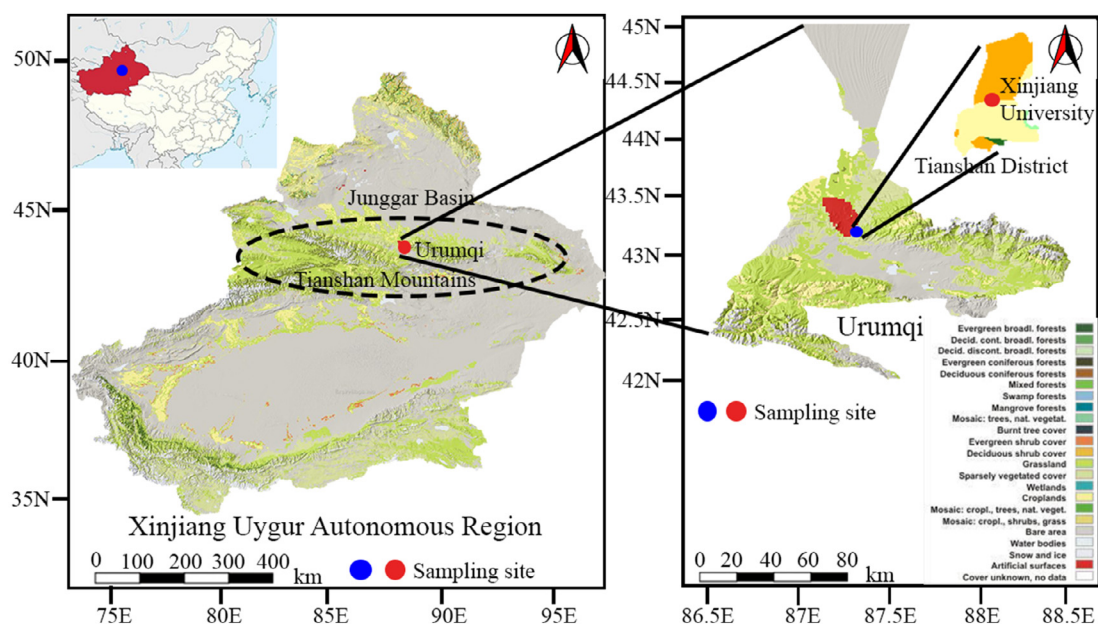


Fig. 1. – Location of sampling site.

$$\begin{aligned}
 &+ 4.8 \times f_L(RH) \times [(NH_4)_2SO_4]_{large} \\
 &+ 2.4 \times f_S(RH) \times [NH_4NO_3]_{small} \\
 &+ 5.1 \times f_L(RH) \times [NH_4NO_3]_{large} \\
 &+ 2.8 \times [OM]_{small} + 6.1 \times [OM]_{large} + [finesoil] \\
 &+ 0.6 \times [coarsematter]
 \end{aligned} \quad (3)$$

where, $[large\ X] = [total\ X]^2 / (20\ \mu g/m^3)$, when $[total\ X] < 20\ \mu g/m^3$;

$[large\ X] = [total\ X]$, when the $[total\ X] > 20\ \mu g/m^3$;

$[small\ X] = [total\ X] - [large\ X]$; here X refers to $(NH_4)_2SO_4$, NH_4NO_3 and OM concentrations.

$[(NH_4)_2SO_4] = 1.375[SO_4^{2-}]$; $[NH_4NO_3] = 1.29[NO_3^-]$ (Dan et al., 2017). Since the metal elements were not measured, fine soil was set to be 20 times Ca^{2+} concentration (Amato and Hopke, 2012). Due to the small contribution of coarse mass to extinction, we ignore them in this study (Hodkinson, 1966). $f_S(RH)$ and $f_L(RH)$ are hygroscopic growth factors (Pitchford et al., 2007). The gas extinction is mainly caused by NO_2 , b_{ag} and b_{sg} (Rayleigh scattering of clear air) are 161 $[NO_2]$ and 10, respectively.

2. Results and discussion

2.1. Characteristic of $PM_{2.5}$ and major species

Fig. 2 shows the concentrations of $PM_{2.5}$, water-soluble ions, OC and EC during the sampling period. $PM_{2.5}$ concentrations in the air of Urumqi largely exceeded the Chinese National air quality standards (daily value of $75\ \mu g/m^3$ and annual value of $35\ \mu g/m^3$). The highest concentrations in fall and spring was $162.75 \pm 80.04\ \mu g/m^3$, and $160.51 \pm 91.46\ \mu g/m^3$, respectively. And the lowest $PM_{2.5}$ levels of $123.43 \pm 36.83\ \mu g/m^3$ occurred in summer. SO_4^{2-} is the most abundant water-soluble ion ($19.64 \pm 20.14\ \mu g/m^3$), accounting for 17.48%–18.84% of $PM_{2.5}$ mass, followed by NO_3^- ($18.25 \pm 19.74\ \mu g/m^3$, 15.69%–16.72%) and NH_4^+ ($12.66 \pm 12.39\ \mu g/m^3$, 11.11%–11.25%). The dominance of the three ions was consistent with the previous results in Urumqi (Chen et al., 2014; Singh et al., 2017). The high

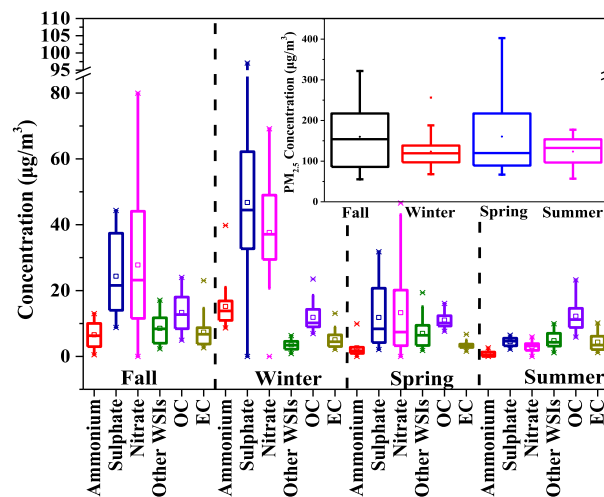


Fig. 2. – Seasonal variation of $PM_{2.5}$, water-soluble ions (WSIs), OC and EC during sampling period (The top down represents the maximum, upper quartile, median, lower quartile, and minimum, and the asterisk represents the outlier).

level of SO_4^{2-} , NO_3^- and NH_4^+ in heating period (November 2017 to March 2018) was likely owing to increase of anthropogenic emissions (i.e. vehicle exhaust and coal combustion) coupled with poor dispersion conditions (such as special topography, low wind speed and high RH). Other water-soluble ions ($5.64 \pm 3.87\ \mu g/m^3$) only contributed 2.54%–5.31% of $PM_{2.5}$ masses. On average, ten water-soluble ions together contributed 37.56% of $PM_{2.5}$ mass. The ten water-soluble ions were sorted by their mass during period of measurement: $SO_4^{2-} > NO_3^- > NH_4^+ > Ca^{2+} > Na^+ > F^- > Cl^- > PO_4^{3-} > K^+ > Mg^{2+}$. The ratio of AE/CE (Anion Equivalent/Cation Equivalent) value slightly lower than one, indicating the $PM_{2.5}$

Table 1. – Average RH and mass concentrations of SO_4^{2-} , NO_3^- and NH_4^+ in different visibility range (Mean \pm SD).

Range of Vis (km)	Average Vis (km)	SO_4^{2-} ($\mu\text{g}/\text{m}^3$)	NO_3^- ($\mu\text{g}/\text{m}^3$)	NH_4^+ ($\mu\text{g}/\text{m}^3$)	Relative Humidity (%)
Vis < 4 (n = 22)	2.03 \pm 0.92	54.64 \pm 36.17	45.26 \pm 30.21	17.04 \pm 12.80	75.96 \pm 8.16
4 < Vis < 7 (n = 25)	5.92 \pm 0.93	17.86 \pm 11.21	21.38 \pm 21.17	4.07 \pm 3.80	46.79 \pm 19.65
7 < Vis < 9 (n = 33)	7.65 \pm 0.34	5.56 \pm 3.03	4.98 \pm 4.97	0.92 \pm 0.82	38.05 \pm 11.03

Table 2. – Correlation matrix of atmospheric visibility verse components in the $\text{PM}_{2.5}$, RH and NO_2 gas concentration during sampling period.

	$\text{PM}_{2.5}$	NH_4^+	NO_3^-	SO_4^{2-}	NO_2	RH
Vis	0.024	−0.676**	−0.670**	−0.724**	−0.663**	−0.781**

** The correlation was significant at the 0.01 level; *: the correlation was significant at the 0.05 level.

aerosols were characterized by alkaline in nature during autumn, spring and summer seasons, while $\text{PM}_{2.5}$ aerosols were neutral in winter (Fig. 1S).

Carbonaceous aerosols were also the main component of $\text{PM}_{2.5}$. Assuming that organic matter (OM) is 1.6 times of OC (Turpin and Lim, 2001), total carbonaceous aerosols (TCA = EC + OM) accounted for 18.46%–20.17% of $\text{PM}_{2.5}$ mass. There was a good correlation between OC and EC, suggesting similar sources (i.e. vehicle exhaust, fossil fuel incomplete combustion and biomass burning (Cao et al., 2005)) (Fig. 2S). OC/EC ratios during this campaign period were in the range of 1.45 to 7.78 with a mean of 2.90. The OC/EC ratios in Urumqi were higher than those reported in the megacities in China (1.68–2.65) (Xiang et al., 2016), indicating substantial secondary organic aerosols in this area (Chow et al., 1994).

2.2. Impact of SNA and RH on visibility

To check the impact of $\text{PM}_{2.5}$ chemical composition on visibility, we categorized the visibility (Vis) data into three groups: Vis < 4 km (low-visibility), 4 km < Vis < 7 km (moderate-visibility), and 7 km < Vis < 9 km (high-visibility). The concentrations of SO_4^{2-} , NO_3^- and NH_4^+ in the low-visibility days were 9.8, 9.1 and 17.4 times higher than those in the high-visibility days, respectively (Table 1). Further analysis found that visibility decreased with the increases of SO_4^{2-} , NO_3^- , NH_4^+ , NO_2 concentrations and RH (Table 2).

Fig. 3 presents the response of visibility with respect to $\text{PM}_{2.5}$, SNA ($\text{SNA} = [\text{SO}_4^{2-}] + [\text{NO}_3^-] + [\text{NH}_4^+]$) and RH. When RH was lower than 70%, the impact of SNA concentrations on visibility was significantly larger than that of $\text{PM}_{2.5}$. And visibility decrease was mainly bond up with SNA concentration. When RH was higher than 70%, visibility reduction was much more dramatic, indicating stronger influence of RH on visibility. Therefore, the attenuated visibility was primary due to the synergistic effect of SNA and RH.

There is a complex nonlinear relationship of visibility with SNA, RH. As Fig. 3 shows, visibility was negatively correlated with SNA. With an increase in the SNA mass concentration, visibility decrease exponentially. We further checked the relationship between visibility and SNA under different RH conditions: RH < 40%, 40% < RH < 70% and RH > 70%. The fitting curves were as follows:

$$\begin{aligned} \text{RH} < 40\%, \text{Vis} &= 9.40 * [\text{SNA}]^{-0.11} \quad (R^2 = 0.45) \\ 40\% < \text{RH} < 70\%, \text{Vis} &= 10.80 * [\text{SNA}]^{-0.17} \quad (R^2 = 0.67) \\ \text{RH} > 70\%, \text{Vis} &= 60.24 * [\text{SNA}]^{-0.67} \quad (R^2 = 0.38) \end{aligned}$$

At the 40% < RH < 70% conditions, visibility exhibited a strong dependence on SNA. And the correlations became weaker when RH was lower or higher. As shown in Fig. 4, the concentration of SNA at 100 $\mu\text{g}/\text{m}^3$ was a critical point. When the concentrations of SNA were higher than 100 $\mu\text{g}/\text{m}^3$, the variation of visibility was not apparent.

2.3. Atmospheric extinction coefficient

Based on the IMPROVE formula, we reconstructed atmospheric light extinction coefficient (b_{ext}) in Urumqi. As shown in Fig. 5, b_{ext} values were the highest in winter ($1441.05 \pm 739.95 \text{ Mm}^{-1}$), and the lowest in summer ($128.58 \pm 58.00 \text{ Mm}^{-1}$). $(\text{NH}_4)_2\text{SO}_4$ and NH_4NO_3 were the major contributors, together accounting for 57.05% of the total b_{ext} . The contributions of carbon component, fine soil and gaseous contaminants to extinction coefficient were 30.19%, 10.3% and 2.46%, respectively (in Fig. 5). High mass fractions of sulfate and nitrate (SNA) in $\text{PM}_{2.5}$ and high RH (>70%) also occurred in winter, resulting in the largest b_{ext} contributions of $(\text{NH}_4)_2\text{SO}_4$ and NH_4NO_3 reaching 87.54%. The proportion of total carbonaceous aerosol was the lowest in winter (10.53%) and the highest in summer (56.41%). The contribution of NO_2 to b_{ext} was the largest in summer. In Urumqi, NO_2 and SO_2 that are precursors of nitrate and sulfate mainly emitted from vehicles exhaust and coal combustion, respectively. In this study, the ratio of $\text{NO}_3^-/\text{SO}_4^{2-}$ range from 0.68 to 1.22 with an average of 0.91, theoretically indicating that the contribution of stationary sources to total light extinction coefficient was slightly higher than that of mobile source during the investigated period.

Light extinction coefficient could be retrieved by Koschmieder formula ($b_{\text{ext-v}} = \text{Ln } \varepsilon / \text{Vis}$, here ε is assumed to be 0.05) (Hinds, 1999). The average values of b_{ext} and $b_{\text{ext-v}}$ were 685.34, 967.92 Mm^{-1} , respectively. Although atmospheric b_{ext} values reconstructed by IMPROVE formula were about 29.19% lower than those by Koschmieder formula, the datasets by both methods were correlated well with each other ($R^2 = 0.87$, Fig. 5). The deviations might be introduced by the several influence factors as follows: 1) The atmospheric pollution in Urumqi is serious (especially the SNA) and high RH during winter (blue box); 2) In warm weather conditions, nitrate will decompose, resulting in the Light extinction coefficient was underestimated via revised IMPROVE formula (red box). In addition, the ε value used in this study might be overestimated since the value in Koschmieder equation would vary from place to place (Jiang et al., 2018).

2.4. Correlation of visibility with light extinction coefficient

There is a strong inverse correlation between b_{ext} and atmospheric visibility. The light scattering coefficients were dominated over light absorption coefficients, since SNA and OM levels were much higher than that EC. Fig. 6 presents the correlations of visibility with light scattering (b_{sp}) and light extinction coefficient (b_{ext}) during the sampling period. The fitting result is completely different from the Koschmieder's

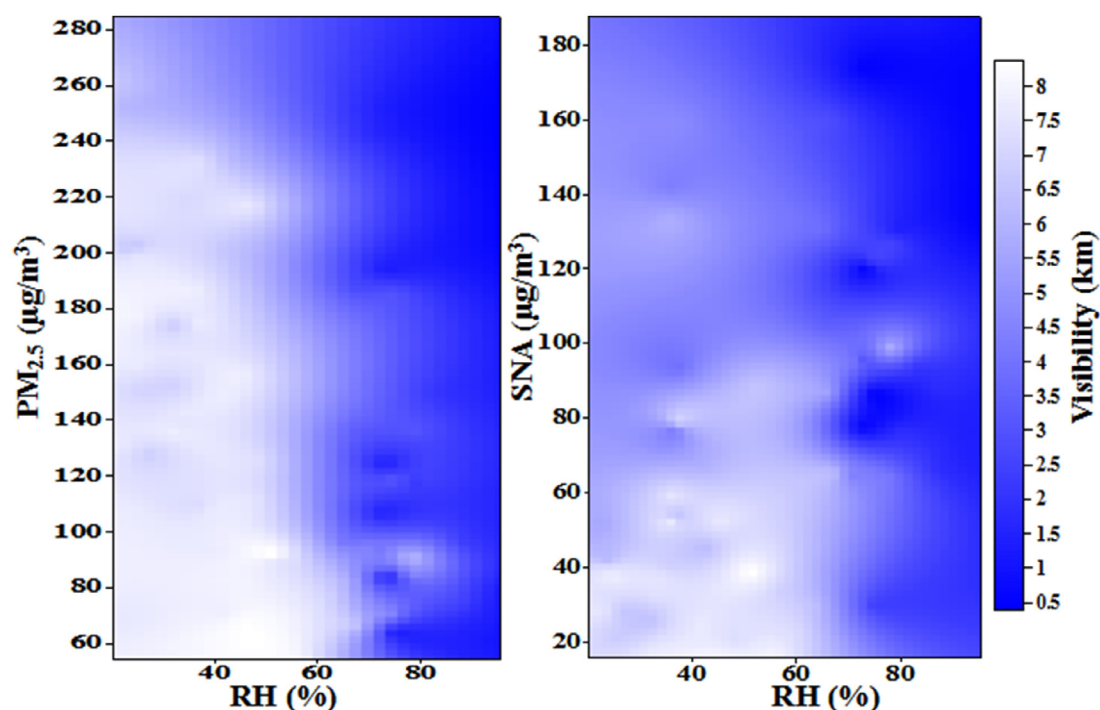


Fig. 3. – Image of mass concentration of PM_{2.5} or SNA (SNA=[SO₄²⁻]+[NO₃⁻]+[NH₄⁺]) concentrations, relative humidity on visibility.

equation in mathematical expression. The inverse relationship between Visibility and b_{sp} or b_{ext} still exists. The power of b_{sp} and b_{ext} in the two fitting formulas are 0.216 and 0.265, respectively, which are much lower than the those of b_{ext} in the Koschmieder's equation. The likely reason may be that the parameter settings in the IMPROVE equation are incompleteness applicable to the southern atmosphere of Urumqi. Another reason is the Koschmieder's equation only applies to ideal observations (Dzubay et al., 1982).

Moreover, the multiple linear regression (MLR) method in R-3.6.0 and SPSS19 was further used to reconstruct the empirical regression model of visibility as a function of (NH₄)₂SO₄, NH₄NO₃, remaining amount of PM_{2.5} (PM_{2.5}-remainder), NO₂ and RH. The results were listed in Table 3. The estimated regression coefficients in the visibility model were examined through t-test, and the *p*-values were less than 0.01 at the 95% confidence level. Although there was a difference between the two equations of linear fitting regression, a good correla-

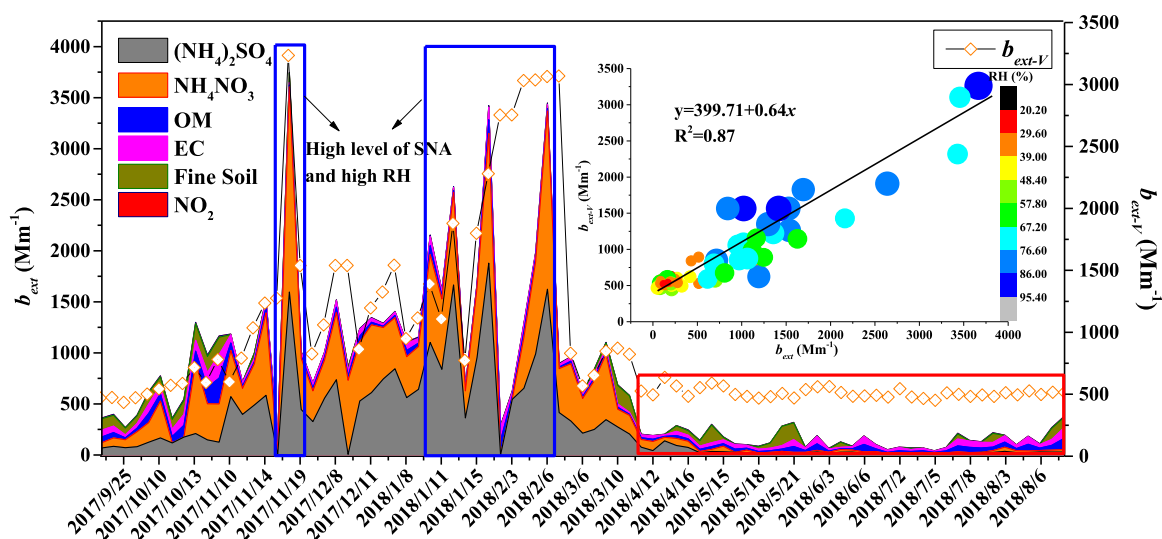
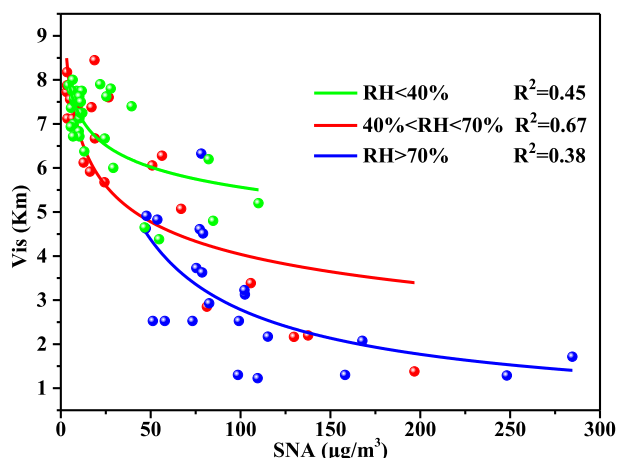


Fig. 5. – Temporal variations of light extinction coefficient via IMPROVE (b_{ext}) and Koschmieder's formulas (b_{ext-v}). The inset illustration represent the comparison of the b_{ext} and b_{ext-v} .

Table 3. – Multiple linear regression results by SPSS and R software for the empirical model of visibility in the Urumqi in sampling period ($Vis = \text{Constant} + a_1 \times [(NH_4)_2SO_4] + a_2 \times [NH_4NO_3] + a_3 \times PM_{2.5}\text{-remainder} + a_4 \times RH + a_5 \times NO_2$).

	Parameter	Estimated regression coefficient (a)	Estimated standard deviation	t value	p value
Eq. (1)(Vis retrieved from MLR with SPSS, $R^2 = 0.822$)	Constant	10.863	0.585	18.573	0.000
	$(NH_4)_2SO_4$	−0.024	0.005	−4.432	0.000
	NH_4NO_3	−0.017	0.006	−2.796	0.007
	$PM_{2.5}\text{-remainder}$	−0.007	0.002	−3.369	0.001
	RH	−0.066	0.009	−7.491	0.000
Eq. (2)(Vis retrieved from MLR with R, $R^2 = 0.807$)	Constant	12.255	0.580	21.117	0.000
	$(NH_4)_2SO_4$	−0.024	0.005	−4.943	0.000
	$PM_{2.5}\text{-remainder}$	−0.008	0.002	−3.706	0.000
	NO_2	−0.028	0.007	−3.801	0.000
	RH	−0.078	0.008	−9.196	0.000

**Fig. 4.** – Scatter plot of the visibility and mass concentration of SNA.

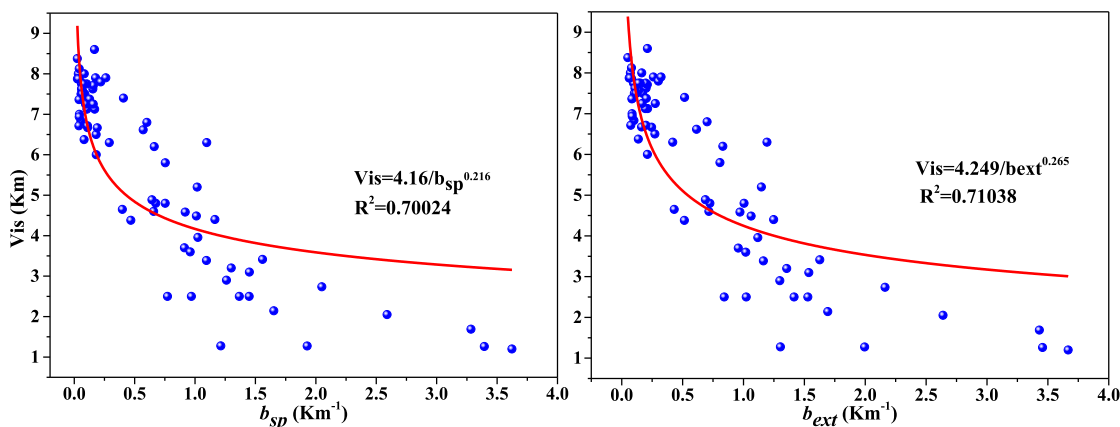
tion was found between the reconstructed visibility and the measured visibility and the slope was close to one (Fig. 3S). This suggested that the two equations could reflect the rela-

tionship between visibility and atmospheric chemical components.

2.5. Principal component analysis of visibility and cases analysis

Atmospheric visibility is directly or indirectly affected by many interactional factors. To identify the major factors controlling the variation of visibility, principal component analysis (PCA) was conducted for the data during the observing period. The data of SO_4^{2-} , NO_3^- , NH_4^+ , Ca^{2+} , OC, EC, RH, wind speed, surface pressure were input in the PCA. As Table 4 showed three main factors were identified by the PCA. All the three factors have statistically significant correlations ($p < 0.05$) with atmospheric visibility.

With the maximum explained variance (58.79%), the factor 1 had strong loading from SO_4^{2-} (0.684), NO_3^- (0.882), NH_4^+ (0.890), RH (0.791), WS (−0.672), pressure (0.843). This factor suggested that a stable and serious air pollution condition could result in low visibility. Low visibility from Nov 13, 2017 to Feb 7, 2018 was affected by this factor (Fig. 5). The second factor, explaining 30.10% of the variance, had high factor loading for OC (0.895) and EC (0.904), indicating that the variation of visibility was related with OC and EC. This factor reflected the influence of carbonaceous aerosols on visibility, especially during Jul 7 to 8 and Aug 1 to 8, 2018. With the explained variance of 11.11%, factor 3 characterized by Ca^{2+} (0.819). During

**Fig. 6.** – the relationship between scattering (b_{sp}) or light extinction coefficient (b_{ext}) and visibility during study period.

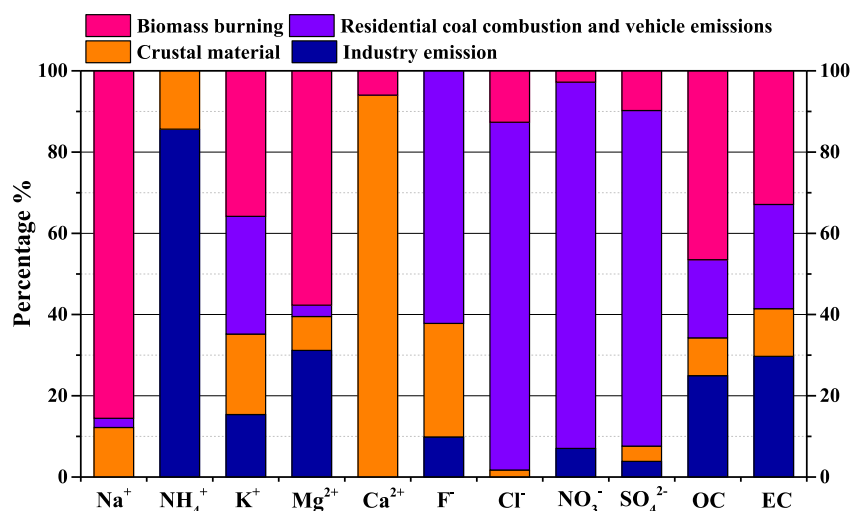


Fig. 7. – Source profile (%) obtained from PMF model analysis.

Table 4. – Factors loading from principal component analysis for the data in Urumqi during sampling period ($n = 80$, values in bold indicate loading factors larger than 0.5).

	Factor 1	Factor 2	Factor 3
SO ₄ ²⁻	0.684	0.360	−0.352
NO ₃ ⁻	0.882	0.121	0.226
NH ₄ ⁺	0.890	0.045	0.036
Ca ²⁺	−0.108	0.311	0.819
OC	−0.006	0.895	0.175
EC	0.146	0.904	0.095
Relative humidity	0.791	−0.230	−0.361
Wind speed	−0.672	−0.170	0.393
Pressure	0.843	0.031	−0.224
Variance (%)	58.79	30.10	11.11

May 18 to 21, 2018, the meteorological conditions were unstable and the sandstorms happened frequently, which significantly affected atmospheric visibility.

2.6. Source contributions to light extinction

Positive matrix factorization (PMF) model has been widely used in source apportionment. We used PMF model to estimate the contributions of aerosol sources that cause visibility deterioration. Data of WSIs (Na⁺, NH₄⁺, K⁺, Mg²⁺, Ca²⁺, F⁻, Cl⁻, NO₃⁻, SO₄²⁻), OC and EC were used as input for the PMF model. Four main sources were identified, including residential coal combustion and vehicle emissions, industrial emission, crustal material and biomass burning (Fig. 7).

Factor 1 is characterized by high loadings of Cl⁻ (85.63%), SO₄²⁻ (82.62%) and NO₃⁻ (90.18%). Previous studies indicated that SO₄²⁻ and NO₃⁻ are formed through the oxidation of SO₂ and NO_x from coal combustion and vehicle exhaust, respectively (Zhang et al., 2013; Tao et al., 2014; Liu et al., 2017). Residential coal burning is an important source of PM_{2.5} in Urumqi (Turap et al., 2019). So factor 1 related to residential coal combustion and vehicle emissions. Due to the lack of other specific indicators (such as V, As and PAHs), PMF cannot separate residential coal burning sources from vehicle exhaust, so treat them as one source. In this study, the contribution of residential coal combustion and vehicle emissions to light ex-

tingtion was 58.6%, probably due to the high emissions of SO₂ and NO_x and strong oxidation in Urumqi during the winter (Zhang et al., 2017).

Factor 2 is dominated by NH₄⁺ (85.64%), Mg²⁺ (31.18%), OC (24.96%) and EC (29.69%), which were closely related to industrial pollution (Wang et al., 2016; D. Wu et al., 2017; Zhou et al., 2016). The contribution of this source to the light extinction was 14.4%. Although Urumqi's air quality has improved in recent years, there are still many industrial zones around Urumqi, i.e. Midong-Wujiaqu industrial zone and coal-fired power plants.

Factor 3 is identified as crustal material on the basis of high leading of Ca²⁺ (94.02%) which are tracer compounds for crustal material (Zhang et al., 2013). The contribution of this source to the light extinction was 9.8%. Crustal material in Urumqi may originate from not only construction and re-suspended road dust, but also transported soil from Gobi Desert and Junggar Basin.

Factor 4 is distinguished by Na⁺ (87.86%), K⁺ (34.19%), Mg²⁺ (52.86%), OC (42.62%) and EC (31.50%) and associated with biomass burning (Zou et al., 2018). The contribution of biomass burning accounted for 17.2% of the light extinction. There are large agricultural areas in Xinjiang. The high contribution of biomass and waste burning should mainly come from the burning of straw during the harvest season.

Thus, residential coal combustion and vehicle emissions was the largest contributor to light extinction in Urumqi, followed by biomass burning, industry emission and crustal material (Fig. 8). Additionally, regional cooperation should be implemented to improve the visibility by reduce emission from all pollution sources.

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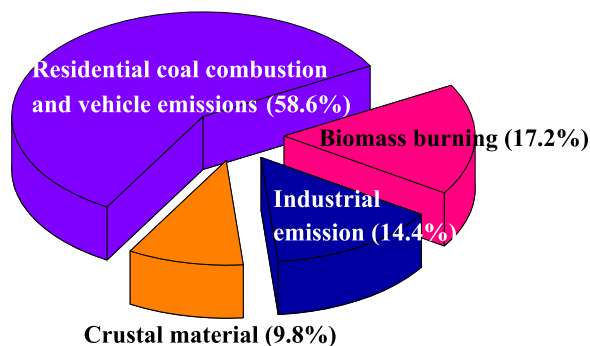


Fig. 8. – Source apportionment of the light extinction in south of Urumqi.

Appendix A. Supplementary data

Supplementary data associated with this article can be found in the online version Journal of Environmental Sciences at doi: 10.1016/j.jes.2020.07.007.

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