Long-term trends in visibility and impacts of aerosol composition on visibility impairment in Baoji, China

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A B S T R A C T

Visibility impairment has become an important environmental issue receiving great attention from both the scientific community and the public. Long-term meteorological data from Baoji, China, were collected to investigate the trend in visibility change from 1980 to 2012. The 33-year average visual range is 12.0 km. The best 20% of the visibility observations in a calendar year shows a general decreasing trend from 1994 onwards, while the worst 20% exhibits a slight increasing trend from 1997 onwards. These results suggest the progressive degradation of air quality in Baoji in recent years. Intensive PM2.5 measurements were conducted from March 2012 to February 2013 to determine the causes of visibility impairment. An analysis based on IMPROVE equation reveals that PM2.5 organic matter (OM) contributes to 34.2% of the light extinction coefficient ($b_{ext}$) on an annual basis, followed by (NH4)2SO4 (30.0%), NH4NO3 (20.1%), elemental carbon (9.2%) and soil dust (6.5%). The largest contributor to $b_{ext}$ for the Worst 20% group is (NH4)2SO4, and the contribution of NH4NO3 for the Worst 20% group increases by a factor of ~3 compared with the Best 20% group. Source apportionment using a positive matrix factorization receptor model indicates that secondary sulfate is the main source of PM2.5 (23.0%), followed by fugitive dust (20.5%), coal combustion (19.9%), secondary nitrate (15.5%), biomass burning (14.3%) and motor vehicle emissions (6.8%). These quantitative results could be useful for policy makers to take effective measures to control the haze pollution in Baoji. Further, the results also are likely to be relevant for other mid-sized cities in China.

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1. Introduction

Visibility impairment in urban environments has attracted considerable attention for scientific community worldwide (Chen and Xie, 2012; Wang et al., 2009; Zhang et al., 2010) and is also a concern for the public, especially those living in urban areas. Visibility degradation is more than an aesthetic issue; in fact, the atmospheric constituents responsible for the problem also affect the human and ecosystem health (Hyslop, 2009). Although natural emissions can be involved, emissions of anthropogenic pollutants are the major cause for degrading atmospheric visibility. Therefore, visibility has commonly been used as an indicator of air quality in urban areas (Watson, 2002). Previous studies (Huang et al., 2009; Thach et al., 2010) have used visibility degradation as a proxy for human exposure...
to study the correlation between air quality and human mortality.

The degradation of visibility is a complicated issue because many factors can affect it simultaneously, including the concentrations, size distributions and composition of aerosol particles, which in turn are subject to the influence of different meteorological conditions (Cao et al., 2012a; Liu et al., 2013; Wang et al., 2013; Tao et al., 2014). Previous studies have shown that elevated concentrations of fine particles are the main cause of visibility impairment, mainly because fine particles scatter visible light more efficiently compared to coarse particles (Deng et al., 2008; Watson, 2002; Yang et al., 2007). In terms of the aerosol composition, sulfate, nitrate, organics and black carbon are the major species that impair visibility (Cao et al., 2012a; Wang et al., 2013; Yuan et al., 2006). Meteorological factors, especially relative humidity, wind speed and the depth of the planetary boundary layer (PBL) influence the concentrations and properties of aerosol particles and thus visibility (Wen and Yeh, 2010; Wu et al., 2005). In addition to the impacts of human activities, several natural phenomena, especially dust storms and fog, can also cause low visibility events (Kim et al., 2001; van Oldenborgh et al., 2010).

Even though particulate pollution has become one of the primary environmental concerns in China, most visibility studies have focused on provincial capitals or developed regions in eastern China (i.e., Che et al., 2009; Tao et al., 2009; Wang et al., 2013). In contrast, there has been relatively little attention paid to the problem in mid-sized cities. The measurement site in our study, Baoji (33°35′–35°06′N, 106°18′–108°03′E), is the second largest city of Shaanxi Province, with a population of over 0.75 million in the downtown area. Baoji is situated in north-central China on the northern bank of the Wei River (Fig. 1). Given that over 40% of the cities in China have a population and gross domestic product in size similar to Baoji (Wu, 2004), our results may have important implications for air pollution in mid-sized Chinese cities.

To better understand the visibility impairment problem in China, it is of particularly advantage to study the past and present visibility. Here, a long-term data set for visual range (VR) was used to investigate the changes in visibility in Baoji from 1980 to 2012. An intensive PM$_{2.5}$ measurement campaign was undertaken to investigate the potential causes of visibility impairment, and receptor modeling was used for source assessments.

2. Data and analysis methods

To investigate the long-term visibility trends in Baoji, China, daily records of VR as well as wind speed and relative humidity (RH) covering the period from 1980 to 2012 were obtained from the Baoji Meteorological Bureau. Well-trained observers measured the VR using reference objects, such as buildings and mountains, in different directions at known distances from the observer (Che et al., 2007). Daily optical light extinction coefficients ($b_{\text{ext}}$) are estimated from the well-known Koschmieder equation (Koschmieder, 1924):

$$b_{\text{ext}} = 3.912/\text{VR}$$

To characterize the impact of PM$_{2.5}$ on visibility impairment, an intensive sampling campaign was conducted on the rooftop (~10 m above ground level) of the Baoji Environmental Monitoring Station (see Fig. 1) from March 2012 to February 2013. Hourly PM$_{2.5}$ loadings were measured with an automatic Environmental Beta Attenuation Monitor (E-BAM, Met One Instruments, Inc., Grants Pass, OR, USA). Meanwhile, 24-hour integrated PM$_{2.5}$ samples were collected every 6 days from...
10 a.m. to 10 a.m. the next day (local standard time) using two battery-powered mini-volume samplers (Airmetrics, Oregon, USA), which operated at a flow rate of 5 L min$^{-1}$.

One sampler was equipped with 47-mm quartz-fiber filters (QM/A; Whatman, Middlesex, UK) for water-soluble inorganic ions, organic carbon (OC) and elemental carbon (EC) analyses, while the other sampler used 47 mm Teflon$^\circledR$ filters (Whatman Limited, Maidstone, UK) for elemental analysis. After sampling, the filter samples were stored in refrigerator at $-4 \, ^\circ\text{C}$ until chemical analysis to prevent evaporation of volatile components. Field blank filters were collected to account for the positive artifacts due to the condensation of gas-phase components. Field blank filters were collected to account for the field blanks. More detailed descriptions of QA/QC procedures can be corrected for the field blanks. More detailed descriptions of QA/QC procedures can be found in Cao et al. (2003).

Seven elements (K, Ti, Mn, Fe, As, Br and Pb) were determined by energy-dispersive X-ray fluorescence (ED-XRF) spectrometry (Epsilon 5 ED-XRF, PANalytical B.V., Netherlands). The Epsilon 5 spectrometer uses a three-dimensional polarizing geometry with 11 secondary targets (i.e., CeO$_2$, CsI, Ag, Mo, Zr, KBr, Ge, Zn, Fe, Ti and Al) and one Barkla target (Al$_2$O$_3$). This configuration provides a good signal-to-background ratio and low detection limits. The X-ray source is a side window X-ray tube with a gadolinium anode, operated at an accelerating power: 600 W). The characteristic X-ray radiation was detected using a liquid nitrogen-cooled, solid-state, germanium, PANalystic (PAN 32) detector.

Water-soluble inorganic ions were determined with a Dionex 600 ion chromatograph ( Dionex Corp, Sunnyvale, CA). Cation (Na$^+$, K$^+$, Mg$^{2+}$, Ca$^{2+}$ and NH$_4^+$) concentrations were separated using a CS12A column ( Dionex Corp.) and 20 mM methane sulfonic acid as the eluent. Anions (SO$_4^{2-}$, NO$_3^-$ and Cl$^-$) were separated with an ASII-HC column (Dionex Corp.) and an eluent of 20 mM KOH. The limits of detection were less than 0.05 mg L$^{-1}$ for both anions and cations. Standard Reference Materials produced by the National Research Center for Certified Reference Materials (China) were analyzed for quality control and assurance purposes. Blank values were subtracted from sample concentrations.

Carbonaceous species (OC and EC) were analyzed by a DRI (Model 2001) Thermal/Optical Carbon Analyzer (Atmoslytic Inc., Calabasa, CA, USA). The method produces data for (A) four OC fractions (OC1, OC2, OC3 and OC4 in a helium atmosphere at 140 °C, 280 °C, 480 °C and 580 °C, respectively); (B) a pyrolyzed carbon fraction (OP, determined when reflected laser light attained its original intensity after oxygen was added to the combustion atmosphere); and (C) three EC fractions (EC1, EC2 and EC3 in a 2% oxygen/98% helium atmosphere at 580 °C, 740 °C and 840 °C, respectively). The IMPROVE A protocol defines OC as OC1 + OC2 + OC3 + OC4 + OP and EC as EC1 + EC2 + EC3 − OP. The analyzer was calibrated with known quantities of CH$_4$ daily. Replicate analyses were performed for one sample in every ten samples. Field blank samples were also analyzed, and the sample results were corrected for the field blanks. More detailed descriptions of Quality Assurance/Quality Control (QA/QC) procedures can be found in Cao et al. (2003).

3. Results and discussion

3.1. Long-term trend in visibility

The mean and median VRs in Baoji are relatively stable from 1980 to 2012 (Fig. 2). The 33-year average VR is 12.0 km, with the maximum annual average occurring in 1995 (13.9 km) and the minimum in 1985 (10.3 km). The 33-year average VR in Baoji is higher than that at Chinese megacities Guangzhou (10.8 km), Beijing (10.7 km), Shanghai (8.6 km) (Chang et al., 2009) or Chengdu (8.5 km) (Wang et al., 2013), but lower than that at Nanjing (16.5 km) and Hangzhou (15.1 km) (Gao et al., 2011).

The 50th percentile (50%) of the observations in a calendar year, the 20th percentile with the lowest VRs (Worst 20%) and the 20th percentile with the highest VRs (Best 20%) are used here as indicators of “median visibility,” “poor visibility” and “good visibility,” respectively, based on the U.S. Regional Haze Rules (Environmental Protection Agency, 1999). As shown in Fig. 2, the Worst 20%, 50% and the Best 20% exhibit similar fluctuations during the study period, with 33-year average values of 4.2, 12.0 and 20.1 km, respectively. For the Worst 20% group, the highest VR is 6.6 km in 2012, and the lowest is 3.1 km in 1997. For the 50% VR group, the highest value is 15.0 km in 2005, and the lowest value is 10.0 km in 1985. For the Best 20% group, the highest VR is 23.2 km in 1989, and the lowest is 16.9 km in 2011. It should be noted that although there is some fluctuation for certain years, the Best 20% of the annual observations show a general decreasing trend of ~2.0 km/decade from 1994 onwards, while the Worst 20% exhibits a slight increasing tendency of ~1.0 km/decade from 1997 onwards. The fluctuation could be associated with the combination effects of PM$_{2.5}$ and meteorological conditions.

Fig. 3a and b show the correlations of the yearly averaged VR with wind speed and RH, respectively. The correlation coefficients show that visual range increase with wind speed, most likely due to enhanced dilution of air pollutants.
under high wind speeds. In contrast, the VR decrease with the increase of RH, most likely because higher RH promotes the growth of hygroscopic particles (e.g., sulfates and nitrates) into large sizes that scatter more light (Watson, 2002).

3.2. Correlation of visual range with PM$_{2.5}$ mass loadings

Fig. 4 shows the daily VR as a function of PM$_{2.5}$ mass concentration during the 1-year intensive study period from March 2012 to February 2013. The data points are color coded for RH and size coded for wind speed. It can be seen that the VR is exponentially related to PM$_{2.5}$ mass, with a correlation coefficient of 0.73. The China Meteorological Administration (2010) defines haze based on two meteorological variables: (i) the atmospheric horizontal visibility is less than 10 km, and (ii) the RH is lower than 80%. In terms of the exponential fit, the threshold of PM$_{2.5}$ mass concentration for the occurrence of haze at Baoji is equivalent to ~112 $\mu$g m$^{-3}$. In other words, when PM$_{2.5}$ mass concentrations are above this threshold value, the visibility in Baoji is typically low. Such analysis therefore provides a scientifically derived PM$_{2.5}$ standard that the local government may use to improve the visibility. It should be noted that this threshold value is 30% higher than the value of 88 $\mu$g m$^{-3}$ in Xi’an (Cao et al., 2012a), but comparable to the value of 110 $\mu$g m$^{-3}$ in Beijing (Zhao et al., 2011). Differences in these local threshold values could be attributed to the different chemical composition of PM$_{2.5}$ in these three cities. Fig. 4 also shows that in general the VR is negatively correlated with RH and positively correlated with the wind speeds during the 2012–2013 intensive campaign period, consistent with the conclusion made from the long-term observations (1980–2012, see Fig. 3).

The correlation matrix in Table 1 illustrates that the VR is negatively correlated with the concentrations of secondary inorganic ions, including SO$_4^{2-}$ ($r = -0.68$), NO$_3^-$ ($r = -0.6$) and NH$_4^+$ ($r = -0.7$). Although the VR is also negatively related to Cl$^-$, Na$^+$, K$^+$, Ca$^{2+}$, OC and EC, the correlation coefficients are much lower, in the range of $-0.5$ to $-0.4$. In addition, NH$_4^+$ is strongly correlated with SO$_4^{2-}$ (0.94) and NO$_3^-$ (0.96), suggesting that these three major ions are in the form of ammonium sulfate and ammonium nitrate. In urban air, the secondary inorganic aerosols are formed through the oxidation of their precursors SO$_2$ and NOx which are mainly emitted from coal burning and vehicle exhausts. Thus, it’s important to control emissions from these two sources.

3.3. Influences of chemical components on light extinction coefficients

The degradation of visibility occurs as a result of the scattering and absorption of light by particles and gases in the
where the quantities in brackets [X] represent the concentrations of the PM$_{2.5}$ components in $\mu$g m$^{-3}$. A RH growth function $f(RH)$ indicates the extent of scattering efficiencies increase for sulfate and nitrate as they absorb liquid water. The $f(RH)$ curve from Malm and Day (2001) is used for this study. The IMPROVE equation has been proven to be a useful tool for assessing the impact of specific types of aerosols on visibility (e.g., Kim et al., 2006; Tao et al., 2009).

Based on the ion-balance calculation (not shown), the SO$_4^{2-}$ and NO$_3^-$ can be completely neutralized by NH$_4^+$ to form (NH$_4$)$_2$SO$_4$ and NH$_4$NO$_3$. Thus, the concentrations of (NH$_4$)$_2$SO$_4$ and NH$_4$NO$_3$ are calculated by multiplying [SO$_4^{2-}$] and [NO$_3^-$] by a factor of 1.40 and 1.29, respectively. The organic matter (OM) and soil dust fractions are estimated from 1.6 × [OC] (Turpin and Lim, 2001) and [Fe]/0.035 (Taylor and McLennan, 1985), respectively. As Baoji is an inland city, the concentration of sea salt could be negligible (Taylor and McLennan, 1985), respectively. As Baoji is an inland city, the concentration of sea salt could be negligible compared to the major composition and was set as zero for simplified purpose. Moreover, the contributions of coarse mass and Rayleigh scattering to $b_{ext}$ have been found to be negligible (Cao et al., 2012a; Cheung et al., 2005), and therefore, these factors also are excluded from the analysis.

As shown in Fig. 5, the reconstructed chemical $b_{ext}$ correlates strongly with the optical $b_{ext}$ estimated from the Koschmieder equation (Koschmieder, 1924). The slope of a least-squares linear regression is 1.09 and $r = 0.78$. These results suggest that the IMPROVE algorithm can provide reasonable estimates for chemical $b_{ext}$ in Baoji under ambient conditions. Nevertheless, at high $b_{ext}$ range, the discrepancies between optical $b_{ext}$ and chemical $b_{ext}$ are larger, which could be associated with the larger uncertainties from the naked-eye observations of lower VR although the observers are well-trained. The uncertainty in $f(RH)$ may also lead to the deviation in estimating the chemical $b_{ext}$. Table 2 shows that the annual average ($\pm$ standard deviation) chemical $b_{ext}$ is $494 \pm 314.8$ Mm$^{-1}$, with the highest value occurring in winter (754 Mm$^{-1}$) followed by autumn (417 Mm$^{-1}$), summer (402 Mm$^{-1}$) and spring (397 Mm$^{-1}$). The annual average chemical $b_{ext}$ at Baoji is higher than what is measured at Jinan (292 Mm$^{-1}$, Yang et al., 2007) or Guangzhou (367 Mm$^{-1}$, Jung et al., 2009), but much lower than that measured at Xi’an (912 Mm$^{-1}$, Cao et al., 2012a) and Chengdu (900 Mm$^{-1}$, Wang et al., 2013).

On average, OM is the largest contributor to the light extinction coefficient, accounting for 34.2% of chemical $b_{ext}$. The fractional contribution of OM to $b_{ext}$ is the highest in winter (41.5%), followed by autumn (35.2%), spring (35.1%) and summer (18.8%). The second most important contributor to $b_{ext}$ is (NH$_4$)$_2$SO$_4$ with an annual average of 30.0%. The highest contribution (48.7%) from (NH$_4$)$_2$SO$_4$ to $b_{ext}$ is found in summer, presumably due to the high concentrations of SO$_4^{2-}$ and high RH compared with the other seasons. The influence of NH$_4$NO$_3$ is also important throughout the year, with seasonal contributions of 16.7–21.8%. Particle light absorption by EC contributes to 9.2% of $b_{ext}$, and the remaining 6.5% of $b_{ext}$ is attributed to soil dust.

Previous studies have shown that the relative influences of chemical species on $b_{ext}$ vary with locations. For urban...
sites in China, Cao et al. (2012a), Yang et al. (2007) and Jung et al. (2009) found that (NH₄)₂SO₄ is the largest contributor to \( b_{\text{ext}} \), accounting for 37–41% of \( b_{\text{ext}} \) in Xi'an, Jinan and Guangzhou, respectively, followed by OM (22–27%). For non-urban sites in USA, Malm and Day (2000) reported that OM contributes to ~35% of \( b_{\text{ext}} \) in the western USA, while sulfate contributes to ~60–70% of \( b_{\text{ext}} \) in the eastern USA. These large differences can be attributed to the variability in the major chemical species responsible for light scattering at different sites as well as to the effects of the ambient RH on particle growth.

To further investigate the contributions of chemical species to \( b_{\text{ext}} \), budgets for the PM₂.⁵ components during the Best 20% and Worst 20% visibility days are calculated. As shown in Fig. 6, for the Best 20% group, OM (39.9%) have the largest effect on \( b_{\text{ext}} \), followed by (NH₄)₂SO₄ (29.5%), EC (12.1%), soil dust (9.7%) and NH₄NO₃ (8.7%). In contrast, for the Worst 20% group, (NH₄)₂SO₄ (32.0%) have the largest effect on \( b_{\text{ext}} \), followed by OM (30.6%), NH₄NO₃ (27.3%), EC (7.1%) and soil dust (3.0%). Even though the contribution of (NH₄)₂SO₄ have the largest effect on \( b_{\text{ext}} \) under the worst VR conditions, it should be noted that the contribution of NH₄NO₃ increased by a factor of ~3 from the Best 20% to Worst 20% group. This indicates that emissions from fossil fuel combustion play an important role in visibility degradation on Worst 20% days.

### 3.4. Mass balance and source apportionment of PM₂.⁵

More than 90% of the PM₂.⁵ mass during the four seasons is constituted by the following six components (Fig. 7): the major water-soluble inorganic ions (SO₄²⁻, NO₃⁻ and NH₄⁺), OM, EC and soil dust. On an annual basis, the most abundant components of PM₂.⁵ are OM (28.0%) and soil dust (27.0%), consistent with the results from fourteen Chinese cities reported by Cao et al. (2012b). During spring, soil dust constitutes the largest fraction of PM₂.⁵, accounting for 42.3% of the mass, followed by OM (21.1%) and SO₄²⁻ (13.6%). In summer, SO₄²⁻ is the largest contributor, accounting for 28.5% of the PM₂.⁵ mass, followed by OM (21.1%), soil dust (20.9%), NO₃⁻ (11.0%), NH₄⁺ (9.6%) and EC (4.4%). During autumn, soil dust (28.2%) and OM (28.0%) are the main components of PM₂.⁵, followed by SO₄²⁻ (14.6%) and NO₃⁻ (12.1%). Finally, in winter, OM is the largest contributor to PM₂.⁵, followed by soil dust (16.6%), SO₄²⁻ (15.5%), NO₃⁻ (13.6%), NH₄⁺ (7.1%) and EC (3.6%).

A positive matrix factorization (PMF) model, which is widely used in source apportionment (Cao et al., 2012a; Chen et al., 2010), is used to estimate the relative contribution of aerosol sources that cause the visibility impairment. The principles of PMF are described in detail elsewhere (Paatero and Tapper, 1994). For the present study, seven elements (K, Ti, Mn, Fe, As, Br and Pb), secondary inorganic ions (NO₃⁻, Table 2

<table>
<thead>
<tr>
<th>Season</th>
<th>(NH₄)₂SO₄</th>
<th>NH₄NO₃</th>
<th>OM</th>
<th>EC</th>
<th>Soil dust</th>
<th>Total ( b_{\text{ext}} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Spring</td>
<td>109.3 ± 104.2</td>
<td>66.1 ± 74.2</td>
<td>139.2 ± 56.8</td>
<td>34.8 ± 16.2</td>
<td>47.2 ± 44.1</td>
<td>396.6 ± 208.3</td>
</tr>
<tr>
<td>% of ( b_{\text{ext}} )</td>
<td>27.6</td>
<td>16.7</td>
<td>35.1</td>
<td>8.8</td>
<td>11.9</td>
<td></td>
</tr>
<tr>
<td>Summer</td>
<td>195.8 ± 176.3</td>
<td>80.9 ± 95.2</td>
<td>75.5 ± 11.2</td>
<td>32.5 ± 11.7</td>
<td>17.1 ± 4.1</td>
<td>401.9 ± 287.6</td>
</tr>
<tr>
<td>% of ( b_{\text{ext}} )</td>
<td>48.7</td>
<td>20.1</td>
<td>18.8</td>
<td>8.1</td>
<td>4.2</td>
<td></td>
</tr>
<tr>
<td>Autumn</td>
<td>103.1 ± 73.6</td>
<td>83.7 ± 67.7</td>
<td>146.8 ± 86.1</td>
<td>51.4 ± 30.4</td>
<td>32.2 ± 23.8</td>
<td>417.0 ± 220.2</td>
</tr>
<tr>
<td>% of ( b_{\text{ext}} )</td>
<td>24.7</td>
<td>20.1</td>
<td>35.2</td>
<td>12.3</td>
<td>7.7</td>
<td></td>
</tr>
<tr>
<td>Winter</td>
<td>181.9 ± 129.0</td>
<td>164.2 ± 149.7</td>
<td>313.0 ± 121.1</td>
<td>62.4 ± 21.5</td>
<td>32.5 ± 14.9</td>
<td>753.9 ± 376.7</td>
</tr>
<tr>
<td>% of ( b_{\text{ext}} )</td>
<td>24.1</td>
<td>21.8</td>
<td>41.5</td>
<td>8.3</td>
<td>4.3</td>
<td></td>
</tr>
<tr>
<td>Annual</td>
<td>148.3 ± 130.8</td>
<td>99.0 ± 107.1</td>
<td>169.0 ± 118.2</td>
<td>45.2 ± 23.9</td>
<td>32.2 ± 27.8</td>
<td>493.6 ± 314.8</td>
</tr>
<tr>
<td>% of ( b_{\text{ext}} )</td>
<td>30.0</td>
<td>20.1</td>
<td>34.2</td>
<td>9.2</td>
<td>6.5</td>
<td></td>
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</table>

Fig. 6. Relative source contributions to chemical light extinction for the Best 20% and the Worst 20% of the annual visual range in Baoji between March 2012 and February 2013. (For interpretation of the references to colour in this figure, the reader is referred to the web version of this article.)
SO$_4^{2−}$ and NH$_4^+$ and carbonaceous species (OC and EC) are used as the model input. The resulting PMF factor profiles are shown in Fig. 8. Linear regression analysis is then used to estimate the contributions of major sources/factors to the PM$_{2.5}$ mass concentrations. The average source contributions calculated are presented in Fig. 9.

Factor 1 is characterized with high load of SO$_4^{2−}$ and NH$_4^+$, which is associated with secondary sulfate and accounts for 23.0% of the PM$_{2.5}$ mass. Factor 2 has high loading of NO$_3^−$ and NH$_4^+$, which is assigned to secondary nitrate. This factor constitutes 15.5% of the PM$_{2.5}$ mass. Factor 3 is characterized with high contents of As, Pb, EC and Mn, most likely associated with coal combustion. This factor accounts for 19.9% of PM$_{2.5}$ mass. Factor 4 has high loadings of Br, EC and OC and is characterized with motor vehicular emissions (Kim and Hopke, 2004). This factor accounts for 6.8% of PM$_{2.5}$ mass. Factor 5 is dominated by crustal elements Ti, Mn and Fe. This factor represents fugitive dust derived mainly from unpaved road dust resuspension and construction activities, which accounts for 20.5% of PM$_{2.5}$ mass.

4. Conclusions

Analyses of a long-term (1980–2012) data set of the visibility in Baoji show that the 33-year average visual range is 12.0 km. Although there is some fluctuation for certain years, the Best 20% of the visibility observations for each year shows a general decreasing trend of ~2.0 km/decade from 1994 onwards, while the Worst 20% of the annual observations exhibits a slight increasing tendency of ~1.0 km/decade from 1997 onwards. This indicates that air pollution in Baoji is getting worse in recent years.

Elevated PM$_{2.5}$ mass concentrations are associated with visibility lower than 10 km, and a PM$_{2.5}$ threshold value of 112 $\mu$m g$^{-3}$ is empirically derived from the relationship between the daily visibility and the PM$_{2.5}$ mass concentration (that is, low visibility is observed when daily PM$_{2.5}$ mass loading exceeds 112 $\mu$m g$^{-3}$). Based on a method established by the IMPROVE program, organic matter (OM) is found to be the largest contributor to visibility impairment overall, accounting for 34.2% of $b_{ext}$, followed by (NH$_4$)$_2$SO$_4$ (30.0%), NH$_4$NO$_3$ (20.1%), EC (9.2%) and soil dust (6.5%). For the Best 20% of the visibility for each year, OM (39.9%) is the largest contributor to $b_{ext}$, followed by (NH$_4$)$_2$SO$_4$ (29.5%), EC (12.1%), soil dust (9.7%) and NH$_4$NO$_3$ (8.7%). In contrast, for the Worst 20% of the visibility, (NH$_4$)$_2$SO$_4$ (32.0%) is the largest contributor to $b_{ext}$, followed by OM (30.6%), NH$_4$NO$_3$ (27.3%), EC (7.1%) and soil dust (3.0%). PM$_{2.5}$ mass balance analysis indicates that OM (28.0%) and soil dust (27.0%) are the most abundant components of the PM$_{2.5}$ mass overall. The PMF receptor model indicates that secondary sulfate is the main contributor to PM$_{2.5}$ mass, accounting for 23.0% of the PM$_{2.5}$ mass loading, followed by fugitive dust (20.5%), coal combustion (19.9%), secondary nitrate (15.5%), biomass burning (14.3%) and motor vehicle emissions. Our results quantify the sources responsible for visibility impairment in Baoji and therefore provide a scientific suggestion for the
development of practical pollution control measures for the local policy makers.

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