Winter brown carbon over six China’s megacities: Light absorption, molecular characterization, and improved source apportionment revealed by multilayer perceptron neural network

Diwei Wang¹, Zhenxing Shen¹*, Qian Zhang², Yali Lei³, Tian Zhang¹, Shasha Huang¹, Jian Sun¹, Hongmei Xu¹, Junji Cao⁴

¹Department of Environmental Science and Engineering, Xi’an Jiaotong University, Xi’an 710049, China
²Key Laboratory of Northwest Resource, Environment and Ecology, MOE, Xi’an University of Architecture and Technology, Xi’an 710055, China
³Key Lab of Geographic Information Science of the Ministry of Education, School of Geographic Sciences, East China Normal University, Shanghai 200241, China
⁴Key Lab of Aerosol Chemistry & Physics, SKLLQG, Institute of Earth Environment, Chinese Academy of Sciences, Xi’an, China

Correspondence to: Zhenxing Shen (zxshen@mail.xjtu.edu.cn)

Abstract. Brown carbon (BrC) constitutes a large fraction of organic carbon and exhibits strong light absorption properties, thus affecting the global radiation budget. In this study, we investigated the light absorption properties, chemical functional bonds, and sources of BrC in six megacities in China, namely Beijing, Harbin, Xi’an, Chengdu, Guangzhou, and Wuhan. The average values of the BrC light absorption coefficient and the mass absorption efficiency at 365 nm in northern cities were higher than those in southern cities by 2.5 and 1.8 times, respectively, demonstrating the occurrence of abundance of BrC in northern China’s megacities. Fourier transform–infrared (FT-IR) spectra revealed sharp and intense peaks at 1640, 1458–1385, and 1090–1030 cm⁻¹, which were ascribed to aromatic phenols, confirming the contribution of primary emission sources (e.g., biomass burning and coal combustion) to BrC. In addition, we noted peaks at 860, 1280–1260, and 1640 cm⁻¹, which were attributed to organonitrate and oxygenated phenolic groups, indicating that secondary BrC also existed in six megacities. Positive matrix factorization (PMF) coupled with multilayer perceptron (MLP) neural network analysis were used to apportion the sources of BrC light absorption. The results showed that primary emissions (e.g., biomass burning, tailpipe emissions, and coal combustion) made a major contribution to BrC in six megacities. However, secondary formation processes made a greater
contribution to light absorption in the southern cities (17.9%–21.2%) than in the northern cities (2.1%–10.2%). These results can provide a basis for the more effective control of BrC to reduce its impacts on regional climates and human health.

1 Introduction

Brown carbon (BrC) constitutes a vital fraction of carbonaceous aerosols and exhibits strong light absorption properties in near-ultraviolet (UV) and visible wavelength regions (Laskin et al., 2015; Wu et al., 2021; Zhang et al., 2022). Therefore, it has received extensive attention in recent years (Laskin et al., 2015; Yan et al., 2018; Yuan et al., 2020). BrC has substantial effects on radiative forcing, cloud condensation, ice cores, and climate (Ma et al., 2020; Sreekanth et al., 2007). On the basis of remote sensing observations and chemical transport model results, studies have detected a BrC-induced nonnegligible positive radiative forcing ranging from 0.1 to 0.6 W m\(^{-2}\) on a global scale (Jo et al., 2016; Wu et al., 2020).

BrC in urban atmospheres can originate from numerous sources, including incomplete combustion of fossil fuels (Soleimanian et al., 2020), biomass burning (Shen et al., 2017; Soleimanian et al., 2020), forest fires, and residential coal combustion (Kirchstetter et al., 2004; Soleimanian et al., 2020). In addition, both primary BrC and gaseous pollutants emitted from anthropogenic and biological activities can be converted into secondary BrC through a series of atmospheric chemical reactions (Kumar et al., 2018; Laskin et al., 2015). Studies have determined that the absorption properties of BrC exhibited distinct temporal and spatial variations in different regions and cities, and these properties were closely related to diverse emissions sources and complex atmospheric aging processes (Chung et al., 2012; Wu et al., 2021). For example, Devi et al. (2016) observed that BrC contributed differently to light absorption in the rural and urban southeast United States. Mo et al. (2021) studied the light absorption coefficient of BrC at 365 nm (BrC \(b_{abs365}\)) in ten Chinese cities, which found that the BrC \(b_{abs365}\) value displayed obvious spatial (northern China > southern China) variations. Furthermore, a stronger light absorption ability in cold seasons (fall and winter) in Beijing (Cheng et al., 2016), Xi’an (Shen et al., 2017), Seoul (Kim et al., 2016), Taiyuan and other cities (Mo et al., 2021) has been found to be strongly associated with increased biomass burning emissions for heating. The mass absorption efficiency at 365 nm (MAE\(_{365}\)) of BrC has been widely used to evaluate the light-absorbing ability of BrC (Bao et al., 2022).
Xie et al. (2017) found that the BrC MAE<sub>365</sub> values from biomass burning (1.28 ± 0.12 m<sup>2</sup> g<sup>-1</sup>) were higher than those from vehicle emissions (0.62 ± 0.76 m<sup>2</sup> g<sup>-1</sup>). Ni et al. (2021) noted that BrC MAE<sub>365</sub> values can be decreased from 1.43 m<sup>2</sup> g<sup>-1</sup> to 0.11 m<sup>2</sup> g<sup>-1</sup> with the BrC aerosol aged. Another study noted that secondary organic aerosol (SOA) formation processes constituted a major source of BrC in Atlanta and Los Angeles; moreover, the optical properties of BrC differed considerably between the two cities due to differences in secondary BrC precursors (Zhang et al., 2011).

China has a high concentration of atmospheric water-soluble organic carbon, which has a major impact on regional air quality, visibility, and the climate (Mo et al., 2021). However, to our knowledge, limited study was conducted to insight to the optical profiles, molecular composition, and sources apportionment of BrC in a large scale in China. Accurately understanding the spatial variations of the sources and light absorption properties of BrC in China is essential for reducing uncertainty about the effects of BrC on the climate. Many studies have used receptor modelling techniques such as positive matrix factorization (PMF) coupled with multiple linear regression analysis to assign the sources of BrC (Bao et al., 2022; Lei et al., 2019; Soleimanian et al., 2020). For example, Bao et al. (2022) obtained specific source contributions to BrC b<sub>abs365</sub> in Nanjing based on PMF and MLR method, confirming that the key contributors to BrC b<sub>abs365</sub> were mainly derived from biomass burning, primary industrial, and traffic emissions. Lei et al. (2018) investigated the source apportionment of BrC b<sub>abs365</sub> in Yulin and showed that the residential coal combustion was the highest contributor to BrC b<sub>abs365</sub> in winter. Soleimanian et al. (2020) used the principal component analysis (PCA) coupled with MLR source apportionment model, which identified fossil fuel combustion was the dominant source of BrC b<sub>abs365</sub> in central Los Angeles during summer (38%), followed by SOA (30%) and biomass burning (12%). However, atmospheric processes are generally non-linear in nature, thus traditional deterministic models could be limited. The artificial neural network (ANN) based models, such as multilayer perceptron (MLP), have been shown to provide meaningful results closer to realistic estimates than most linear models (Borlaza et al., 2021a; Elangasinghe et al., 2014). Therefore, in this study, a winter campaign for PM<sub>2.5</sub> sampling was conducted over six China’s megacities. The purposes of this study were to 1) investigate the spatial variations of the carbonaceous matter concentrations and optical properties of BrC across six representative urban areas in China, 2) determine the molecular composition of BrC, and 3) insight the relationship between light absorption and BrC sources by using PMF coupled with ANN-MLP.
2 Methods

2.1 Samples collection

PM$_{2.5}$ samples were collected in six cities in China (Figure 1): three cities in northern China (Beijing [BJ], Harbin [HrB], and Xi’an [XA]) and three cities in southern China (Chengdu [CD], Guangzhou [GZ], and Wuhan [WH]). We classified the cities as being in northern or southern cities according to their geographic location, such as “north or south of the Huaihe River”. Owing to geographical factors, these cities exhibit considerable differences in terms of energy structure and climate. The average annual temperature in northern cities is generally below 15°C, while in southern cities it is usually above 15°C (Mo et al., 2021). Information about the six cities and the sampling sites is summarized in Table S1 (Supporting Information).

Figure 1. PM$_{2.5}$ samples were taken in six Chinese cities.

For sample collection, filter samplers were mounted on rooftops between 8 and 30 m above the ground, and samples were collected from November 20 to December 22, 2019. In BJ, HrB, and GZ, a mini-volume sampler operating at 5 L min$^{-1}$ (Airmetrics, Springfield, OR, USA) was used to collect PM$_{2.5}$ samples on 47-mm quartz-fiber filters (Whatman, Maidstone, UK) for 24 h. In CD, a medium-volume PM$_{2.5}$ sampler operating at 100 L min$^{-1}$ (HY-100SFB, Hengyuan, Qingdao, China) was used to collect PM$_{2.5}$ samples on 90-mm quartz-fiber filters (Whatman). Moreover, in XA and WH, a high-volume sampler (HVS-PM$_{2.5}$, Thermo-Anderson Inc. Cleves, OH, USA) with a flow rate of 1.13 m$^3$ min$^{-1}$ was used to collect PM$_{2.5}$ samples on quartz-fiber filters (203 mm × 254 mm, Whatman, QMA). Before sample collection, all quartz filters were prebaked at 780 °C for 7 h to eliminate any residual carbon. A
detailed description of the quality control procedures for the filters before and after the sampling processes can be found in the article by Shen et al (2017). After the sampling processes, the samples were sealed and stored below 0 °C to avoid evaporative losses before analysis.

2.2 Chemical analysis

The organic carbon (OC) and elemental carbon (EC) of the PM$_{2.5}$ samples were analyzed using a Thermal and Optical Carbon Analyzer (DRI Model 2001A, Atmoslytic, Inc., USA) in accordance with the improved Interagency Monitoring of Protected Visual Environment (IMPROVE) thermal/optical reflectance protocol. Detailed descriptions of the OC and EC measurement methods can be found in the article by Cao et al (2004). A portion of each filter (about 2.84 cm$^2$) was extracted using 10 mL of ultrapure water to analyze water-soluble inorganic ions (Na$^+$, NH$_4^+$, K$^+$, Mg$^{2+}$, Ca$^{2+}$, Cl$^-$, NO$_3^-$, and SO$_4^{2-}$) through ion chromatography (Dionex 500, Dionex Corp, USA). A detailed description of the ion analysis method used in this study can be found in the article by Shen et al (2008).

2.3 Optical properties of methanol extracts

A 0.526-cm$^2$ punch was ultrasonically extracted from each filter sample by using 5 mL of methanol (HPLC Grade, Fisher Scientific, NH, USA) for 30 min. Subsequently, all extracts were filtered through a microporous membrane with a diameter of 25 mm and pore size of 0.22 μm (Puradisc 25 TF, PTFE membrane) to remove insoluble components. The UV–visible absorption spectra of the BrC samples were determined using a liquid waveguide capillary cell–total OC spectrophotometer (LWCC-2100, World Precision, Sarasota, FL, USA) between the wavelengths of 200 and 700 nm. The BrC optical properties such as $b_{abs365, methanol}$ (The absorption coefficient for methanol exacts at 365 nm) and $MAE_{365, methanol}$ (normalized by $b_{abs365, methanol}$ to organic carbon, OC) were calculated as showed in previous study (Lei et al., 2019) and details was listed in Text S1.

2.4 Fourier transform infrared spectroscopy spectra

Functional groups in the samples collected in six megacities were characterized using a Fourier transform infrared (FT-IR) spectrometer (Bruker Optics, Billerica, MA, USA). The method described in section 2.3 was used to extract the BrC filtrates, then the BrC extracts were concentrated to 0.5 mL under a gentle nitrogen flow, after which they were mixed with 0.2 g of KBr (FT-IR grade, Sigma-Aldrich) and
then blown with nitrogen to complete dryness. The resulting extract–potassium bromide mixture was
ground in an agate mortar and examined through FT-IR spectroscopy. The FT-IR spectrum of each sample
was recorded in transmission mode by averaging 64 scans using a standard optical system with KBr
windows. The spectra were recorded in the wavelength range of 4000–400 cm$^{-1}$ at a resolution of 4 cm$^{-1}$.

Before analyzing the aerosol extract samples, we obtained the baseline spectrum by analyzing pure KBr.

### 2.5 Source apportionment of BrC light absorption coefficient at 365 nm

In this study, the source apportionment of BrC was conducted using the PMF coupled with ANN-MLP
methods by following the steps: 1) identification and quantification of the major sources of PM$_{2.5}$ for the
six cities using PMF (The United States Environmental Protection Agency, PMF 5.0); 2) produces a
predictive model by ANN-MLP for one variable (BrC $b_{abs365}$) based on the values of the input variables
(PM$_{2.5}$ sources daily contributions). PMF is a bilinear factor model that has been widely used in source
apportionment studies (Cao et al., 2012; Lei et al., 2018; Li et al., 2021; Shen et al., 2010; Tao et al.,
2017). In the present study, water-soluble inorganic ions (Na$^+$, NH$_4^+$, K$^+$, Mg$^{2+}$, Ca$^{2+}$, NO$_3^-$, SO$_4^{2-}$ and
Cl$^-$) and carbon fractions (OC1, OC2, OC3, OC4, EC1, and EC2) were used as data inputs for PMF. The
PMF model was run multiple times, extracting four to six factors. A more detailed description of these
items can be found in the article by Lei et al (2019). Subsequently, an MLP model was constructed. The
model was developed using IBM SPSS Statistics for Windows, version 23 (IBM Corp., Armonk, NY,
USA). The detail information of the ANN-MLP model construction and training was described in Text
S2. After ANN-MLP model training, the obtained MLP model was applied to a set of virtual datasets.
Each virtual dataset consists of each source with the same mass contribution (from PMF analysis) as the
original dataset, but with one source set to zero. The BrC $b_{abs365}$ contribution for a specific source was
obtained by subtracting the BrC $b_{abs365}$ simulation value obtained using the virtual dataset from the BrC
$b_{abs365}$ simulation value obtained using the original MLP model, which contains all the source
contributions (Borlaza et al., 2021a).

### 3 Results and discussion

#### 3.1 General description of PM$_{2.5}$ and its chemical species in six megacities

As presented in Table S2, the PM$_{2.5}$ concentrations in six cities ranged from 9.9 to 241.9 μg m$^{-3}$ and
exhibited a significant spatial variation ($p < 0.01$), indicating the complexity of air pollution and spatial
differences in air pollution levels in China. HrB had the highest average PM$_{2.5}$ concentration (85.5 ± 43.9 μg m$^{-3}$), which exceeded National Air Quality Standard grade-II (24-h average: 75 μg m$^{-3}$) and was 1.1, 1.2, 2.0 and 1.3 times higher than those recorded in BJ, XA, CD, GZ, and WH, respectively. This phenomenon indicates that PM$_{2.5}$ pollution is still a major challenge in China, particularly in northern China.

The average concentration of OC, a major chemical component of PM$_{2.5}$, ranged from 5.6 to 19.4 μg m$^{-3}$ in six megacities; these cities can be arranged (in descending order) as follows in terms of the average OC concentration: HrB > XA > BJ > WH > GZ > CD (Table S2). Similar to the PM$_{2.5}$ trend, the average OC concentration in the northern cities (15.5 ± 7.9 μg m$^{-3}$) was higher than that in the southern cities (9.2 ± 4.6 μg m$^{-3}$), which can be attributed to substantial emissions from residential heating (i.e., coal and biomass combustion) in winter in northern China (Zhang et al., 2021). In addition, these residential fuels can emit an abundant OC emission (Lei et al., 2018; Sun et al., 2017). To assess the sources of atmospheric BrC, we estimated the concentrations of primary OC (POC) and secondary OC (SOC) by using the EC tracer method (Ram and Sarin, 2011). Detailed calculation method was described in Text S3. As presented in Table S2, the average SOC concentrations throughout the measurement period ranged from 1.0 (CD) to 9.2 μg m$^{-3}$ (HrB), and the fractional contributions of SOC to OC varied from 22.6% to 66.6%. The average POC concentrations ranged from 4.0 (GZ) to 10.2 μg m$^{-3}$ (HrB), and POC constituted 34.4%–77.4% of the total OC mass in the six cities. Accordingly, the SOC and POC concentrations exhibited typical spatial fluctuations, which were consistent with the fluctuations of the PM$_{2.5}$ and total OC concentrations. These results reveal that primary emissions usually dominated secondary formation processes, especially in the northern cities.

3.2 Light absorption properties of BrC

As plotted in Figure 2, the light absorption coefficient ($b_{abs}$, Mm$^{-1}$) values for BrC exhibited significant spatial variations across the six cities (1.7–64.1 Mm$^{-1}$; $p < 0.01$). We executed Student $t$ test at the 95% confidence level and observed that HrB had the highest average BrC $b_{abs365}$ value (29.3 ± 14.2 Mm$^{-1}$), followed by BJ (11.4 ± 3.9 Mm$^{-1}$), WH (10.0 ± 3.2 Mm$^{-1}$), XA (8.3 ± 2.4 Mm$^{-1}$), CD (5.6 ± 2.7 Mm$^{-1}$), and GZ (4.3 ± 1.4 Mm$^{-1}$). The average BrC $b_{abs365}$ value in the northern cities was 15.7 ± 12.3 Mm$^{-1}$, which was 2.5 times higher than that in the southern cities ($p < 0.01$). The large variation in the measured BrC $b_{abs365}$ values in these megacities was observed, which reflected that the light absorption of BrC was
heavily affected by chromophore sources (Huang et al., 2018; Soleimanian et al., 2020), aging during atmospheric transportation (Lambe et al., 2013), and meteorological conditions (Li et al., 2021). Light-absorbing carbonaceous aerosols were believed to be responsible for the considerable absorption of light in the atmosphere (Xie et al., 2020). As presented in Figure S2, we observed positive correlations between BrC $b_{abs365}$ and POC in the six cities ($r$ range: 0.61–0.92). Similar correlations were observed between BrC $b_{abs365}$ and SOC ($r$ range: 0.51–0.80), indicating that the sources of atmospheric BrC in the six cities were quite complex. Apart from primary emissions, secondary formation processes also seemed to have a considerable contribution to BrC in these cities. Biomass burning was revealed to be the dominant source of BrC in these cities during winter (Cheng et al., 2016; Shen et al., 2017; Sun et al., 2017; Cheng et al., 2022). Furthermore, we observed high correlations ($r$ range: 0.69–0.92) between BrC $b_{abs365}$ and K$^+$, which is commonly regarded as a tracer of biomass burning (Shen et al., 2010), in HrB, BJ, XA, and WH (Figure S3). This evidence supports the aforementioned findings that emissions from biomass burning might be the major BrC source in winter in these cities. For the southern cities CD and GZ, the low BrC $b_{abs365}$ values (1.7–11.5 Mm$^{-1}$) are of the same order of magnitude as those reported previously in Nanjing (3.3–13 Mm$^{-1}$; Chen et al., 2019; Chen et al., 2018), Seoul (0.9–7.3 Mm$^{-1}$; Kim et al., 2016), and Hong Kong (4.8–10.6 Mm$^{-1}$; Zhang et al., 2020). The aging or oxidation of aerosols were confirmed to be the major source of BrC in these regions, indicating that secondary aerosols are likely a major source of winter BrC in CD and GZ.
Figure 2. Spatial variations of BrC light absorption properties from six Chinese cities. The bars represent the light absorption coefficient at 365 nm ($b_{abs365}$, left axis), and the lines represent the mass absorption efficiency at 365 nm (MAE$_{365}$, right axis).

The mass absorption efficiency (MAE, m$^2$ g$^{-1}$) is a key parameter for describing the light absorption ability of atmospheric BrC (Li et al., 2021; Peng et al., 2020). Figure 2 illustrated the average MAE values measured at 365 nm (BrC MAE$_{365}$) in the six cities; compared with the value measured in CD (0.37 ± 0.18 m$^2$ g$^{-1}$), those measured in the other five cities were higher by 1.1–3.3 times. These cities can be arranged as follows (in descending order) in terms of the measured BrC MAE$_{365}$ values: HrB > BJ > WH > XA > GZ > CD. These differences in BrC MAE$_{365}$ values can be attributed to the variance of the light absorption capacity of BrC in different megacities. The average BrC MAE$_{365}$ values measured in BJ, HrB, XA, and WH (range: 0.68–1.21 m$^2$ g$^{-1}$) were within the MAE ranges of biomass burning, such as, the average MAE$_{365}$ measured for BrC were 0.97 ± 0.26 m$^2$ g$^{-1}$ for wood burning (Du et al., 2014), 1.05 ± 0.08 m$^2$ g$^{-1}$ for corn stalk combustion (Du et al., 2014), and 1.28 ± 0.12 m$^2$ g$^{-1}$ for wheat stubble burning (Xie et al., 2017; Lei et al., 2018), indicating that biomass burning may be a major source of winter BrC in these cities. Biomass burning is commonly regarded as the main emission source for BrC, which has a high absorption capacity, as indicated by field observations and model predictions (Desyaterik et al., 2013; Feng et al., 2013; Lei et al., 2018). Notably, the MAE$_{365}$ values derived for BrC emitted from primary fossil fuel combustion are similar to those derived for biomass burning (Yan et al., 2017); for example, former studies have revealed that the BrC MAE$_{365}$ values produced by primary emissions from residential coal combustion were in the range of 0.30–1.51 m$^2$ g$^{-1}$ (Ni et al., 2021; Yan et al., 2017). Therefore, coal combustion may also be a potential source of BrC in these cities. By contrast, we observed lower average BrC MAE$_{365}$ values in GZ and CD (range: 0.37–0.39 m$^2$ g$^{-1}$). Previous studies have revealed relatively low BrC MAE values from motor vehicle emissions, including gasoline vehicle emissions (0.62 ± 0.76 m$^2$ g$^{-1}$; Xie et al., 2017) and motorcycle emissions (0.20 ± 0.08 m$^2$ g$^{-1}$; Du et al., 2014). These findings suggest that the BrC sampled in GZ and CD mainly originated from traffic emissions. In addition, laboratory experiments in a previous study revealed that MAE$_{365}$ values decreased from 1.43 to 0.11 m$^2$ g$^{-1}$ with aerosol aging, which suggests the production of SOA (Ni et al., 2021). This finding demonstrates that secondary formation processes are among the main sources of BrC in CD and GZ.

The absorption Ångström exponent (AAE) measurements at 330–550 nm represents the wavelength
dependence of light absorption by BrC (Cheng et al., 2017). We observed that the average AAE values for BrC varied from 5.4 to 6.8 in the six cities (Figure 3). In general, the AAE values obtained in this study were higher than those obtained at the Nepal Climate Observatory-Pyramid (3.7–4.0; 330–500 nm) (Kirillova et al., 2016) and in the Los Angeles Basin (4.82 ± 0.49; 300–600 nm) (Zhang et al., 2013) and lower than those obtained at the Tibetan Plateau (8.2 ± 1.4; 365–550 nm) (Zhu et al., 2018). Nevertheless, the values obtained in this study were comparable to those obtained in Beijing (5.3–7.3; 310–450 nm) (Cheng et al., 2016; Wu et al., 2021), Nanjing (6.7; 300–600 nm) (Chen et al., 2018), the Indo-Gangetic Plain (5.3; 300–700 nm) (Srinivas et al., 2016), New Delhi (5.1; 330–400 nm) (Kirillova et al., 2014), Seoul (5.5–5.8; 300–700 nm) (Kim et al., 2016), and Xi’an (5.3–6.1; 330–550 nm) (Huang et al., 2018). These similarities can primarily be attributed to the consistent solubility of chromophores, which are sensitive to the type of fuel used, the combustion conditions, and the solvents used (Cao et al., 2021; Huo et al., 2018). Furthermore, the AAE values obtained in this study were within the range of those reported by previous studies for coal combustion (5.5–6.4; 300–500 nm) (Ni et al., 2021), biomass burning (4.4–8.7; 300–550 nm) (Xie et al., 2017), and gasoline vehicle emissions (6.2–6.9; 300–550 nm) (Xie et al., 2017). This suggested that BrC in our study may have multiple sources. Additionally, in contrast to the trends observed for the BrC \text{abs}_{365} \text{ and BrC MAE}_{365} values in the various cities, the AAE values observed in CD and GZ were higher than those observed in the other cities. A previous study reported that the AAE values for SOA were higher than those for primary organic aerosols (Saleh et al., 2013), and previous laboratory combustion experiments revealed that the aging of biomass burning aerosols generally engenders an increase in AAE values (from 6.93 to 15.59; Sengupta et al., 2018). These findings suggested that BrC in the cities in this study was also affected by secondary formation processes.
Figure 3. AAE values of BrC in six cities. AAE is calculated between 330 and 550 nm.

3.3 Molecular structure of BrC

In order to further explore the reasons for the differences in the optical properties of BrC among these cities, the functional groups of BrC were measured using FT-IR spectroscopy. Figure 4 illustrates the FT-IR spectra of BrC fractions within the region of 4000–400 cm\(^{-1}\) in the six cities. The band in the region of 400–800 cm\(^{-1}\) resulted from the interference from water vapor inside the instrument and thus can be ignored (Zhang et al., 2020). The broad and strong peak at 3450 cm\(^{-1}\) was contributed to the O-H stretch of H-bonded hydroxyl groups, phenols and carboxylic (Fan et al., 2016; Mukherjee et al., 2020). The sharp band near 1740 cm\(^{-1}\) was usually assigned to the C=O bonds of ketones, quinones, and amides (Duarte et al., 2005; Kristensen et al., 2015). We also attributed the sharp and intense absorption peaks at 2850–2990 cm\(^{-1}\) to aliphatic asymmetric and symmetric C–H stretching vibrations (Coury and Dillner, 2008). Some bands were also displayed near 1640, 1458 and 1030 cm\(^{-1}\), previous studies confirmed that these bands were generally ascribed to the C=O bonds of ketones, quinones, and amides (Fan et al., 2016; Zhao et al., 2022), indicating the presence of aromatic groups. These results demonstrate the complexity of the chemical composition of BrC in the six cities, mainly containing aliphatic chains, carboxylic groups, and aromatic groups.

In contrast to these similar functional groups, the apparent differences of typical functional bands were
also found among these cities. The strong band near 3130 cm$^{-1}$ denoting O–H band (Fan et al., 2016; Mukherjee et al., 2020) were only detected in XA, CD and WH, and the same peak were observed in the spectra from the corn straw burning (Fan et al., 2016) and coal combustion (Zhang et al., 2022), which stressed the emissions of biomass burning and coal combustion with high abundance of oxygenated phenolic compounds in these cities. Moreover, the peak at 1385 cm$^{-1}$ was generally considered to be derived from the O–H bond deformation and C–O stretching of phenolic groups (Fan et al., 2016; Mukherjee et al., 2020; Zhang et al., 2020), and the same peak was observed in the FT-IR spectra of BrC samples derived from the combustion of biomass materials (Fan et al., 2016). These observations indicated the contribution of biomass burning to BrC; this was because that biomass burning can release heat-modified lignin derivatives such as aromatic phenols (e.g., syringyl and guaiacyl) (Duarte et al., 2007; Fan et al., 2016; Zhao et al., 2022). It was noted that the abundance of this peak was different among six cities, and was significantly higher in HrB, XA and WH, which indicated biomass burning contributed differently to BrC in six cities, and higher contribution was occur in HrB, XA and WH than those in other cities. Previous studies have shown that BrC from biomass burning has a high light absorption capacity (Cao et al., 2021; Desyaterik et al., 2013; Kumar et al., 2018), which supported that these cities with higher abundance of aromatic phenol functional groups were consisted with higher BrC $b_{abs365}$ (range: 8.3–29.3 Mm$^{-1}$) and BrC MAE$_{365}$ (range: 0.68–1.21 m$^2$ g$^{-1}$) values in section 3.2. Furthermore, we observed three peaks at 860, 1280–1260, and 1640 cm$^{-1}$, demonstrating the presence of organic-nitrate (C–ONO$_2$) and oxygenated phenolic groups (Day et al., 2010; Zhang et al., 2020). Previous studies have shown that the anthropogenic volatile organic compounds, sulfates, nitrates and other acidic particle components from coal and biomass combustion may enhance the contents of these functional groups through aqueous-phase formation under high humidity conditions (Gilardoni et al., 2016; Wang et al., 2019; Zhang et al., 2020). Therefore, the FT-IR spectra indicated that all the BrC samples from six cities have the contribution of secondary generation. Besides, the abundance of functional groups at these wavenumbers, especially at 1640 cm$^{-1}$, was higher in CD than that in other cities. These results might indicate that the secondary source of BrC was relatively high in CD.
3.4 Source apportionment of BrC

Considering the complexity of atmospheric processes, and the correlation and/or nonlinear interaction between independent variables (i.e., multicomponent or multi-source interactions), we attempted to apply ANN techniques of nonlinear functions, such as MLP model, combined PMF analysis to predict the source contribution of allocated BrC from PM$_{2.5}$ sources in this study. The PMF-apportioned source contributions to PM$_{2.5}$ in the six cities are presented in Figures S4 and S5. A good correlation was observed between the measured and PMF-reconstructed PM$_{2.5}$ mass concentrations in all sites (BJ: $r = 0.99$; HrB: $r = 0.90$; XA: $r = 0.97$; CD: $r = 0.97$; GZ: $r = 0.94$; WH: $r = 0.95$). Theoretical $Q_{\text{true}}$ and $Q_{\text{robust}}$ displayed a <5% difference, and scaled residuals of >95% data were in the range of −3 to 3. These evidences demonstrated the validity and robustness of our PMF solutions (Borlaza et al., 2021b; Tao et al., 2021). As illustrated in Figure S4, the first source was dominated by sulfate, OC, and EC and was considered to represent from coal combustion (Huang et al., 2014). The second source comprised high concentrations of NH$_4^+$, NO$_3^−$, and SO$_4^{2−}$ and was considered to represent secondary formation processes (Shen et al., 2010). Furthermore, the third source comprised high loadings of K$^+$ and was considered to represent biomass burning (Shen et al., 2010). The fourth source primarily comprised Na$^+$, Mg$^{2+}$, and Ca$^{2+}$ and was thus determined to represent fugitive dust (Shakeri et al., 2016; Shen et al., 2016; Sun et
The fifth source contained high concentrations of Mg$^{2+}$, Ca$^{2+}$, NO$_3^-$, OC, and EC and was thus identified as representing traffic-related emissions (Shakeri et al., 2016). Finally, the sixth source comprised high concentrations of OC, EC, and NO$_3^-$ and was considered to represent vehicle emissions (Shakeri et al., 2016).

The optimal neural network model for each site were explored by changing activation function types (Tan H and Sigmoid), optimizing algorithms (scaled conjugate and gradient descent), and based on the lowest root mean square error (RMSE) and the highest correlation coefficient ($r$) between observed and MLP-modelled values (Borlaza et al., 2021a). Although there are other architectures that are more complex for MLP models, a basic MLP architecture was considered sufficient for the input and output data sets of this study.

Figure S6 shows the correlation between observed values and BrC $b_{abs365}$ predicted values from selected MLP models. The good correlation indicated the reliability of the model results. On the basis of the MLP results, we calculated the source-specific contributions to BrC in the six cities (Figure 5). The primary sources including coal combustion, dust, vehicle, biomass burning and traffic emissions, and their average contribution to BrC in the northern cities was 93.3%, which was 1.2 times higher than that in the southern cities. Among these primary emissions, we noted that a higher contribution of biomass burning to BrC in HrB, BJ, XA and WH compared to other cities, which is consistent with the higher abundance of biomass burning products, such as aromatic phenol functional groups was founded in these cities as discussed in section 3.3. As supported, the BrC from biomass burning have high MAE$_{365}$ values (Cao et al., 2021; Kumar et al., 2018), which can be also observed among these cities (range: 0.68–1.20 m$^2$ g$^{-1}$). In addition, we noted that the contribution of biomass burning to BrC in WH (37.7%) was higher than that in CD (13.6%) and GZ (0%), which can explain the highest BrC MAE$_{365}$ was observed in WH among southern cities as shown in Figure 2. On average, the secondary formation source contribution to BrC in southern cities was 19.4%, which was 2.9 times higher than that in northern cities. Besides, the highest contribution was observed in CD with 21.2%, followed by GZ > WH > BJ > HrB > XA. This result can be supported by the abundance of organic-nitrate functional groups, the relatively high AAE value and low BrC MAE$_{365}$ value in CD, which were closely related with the contribution of secondary sources.
Figure 5. The source contribution to BrC using multilayer perceptron neural network analysis in (a) BJ, (b) HrB, (c) XA, (d) CD, (e) GZ, (f) WH.

4 Conclusions

We investigated the sources and light absorption properties of BrC in wintertime in six megacities across China. Both the \( b_{abs} \) and MAE\(_{365} \) of BrC at 365 nm in northern cities were approximately 2.5 and 1.8 times higher than those in southern cities. The BrC MAE\(_{365} \) values measured in BJ, HrB, XA and WH were ranged from 0.68 to 1.21 m\(^2\) g\(^{-1}\), which were within the MAE ranges derived for biomass burning. Thus, these comparisons confirmed that emissions from biomass burning might be the major BrC source in winter in these cities. Previous studies have reported that MAE\(_{365} \) values decreased with aerosol aging while the AAE values of SOA were higher than those for POA. Besides, we noticed that the average BrC MAE\(_{365} \) and AAE values showed different trends in southern cities of CD and GZ, that is, the BrC MAE\(_{365} \) values of these two cities were lower than those of other cities, while the AAE values were relatively higher. These evidences supported the secondary formation process were among the main sources of BrC in CD and GZ.

The chemical functional groups of BrC in six cities mainly included aliphatic chains, carboxyl groups and aromatic groups. However, the apparent difference of typical functional bands revealed the important...
contributions of primary biomass burning and coal combustion to BrC for high abundance of oxygenated phenolic compounds in these cities, especially in HrB, XA and WH. In contrast, the presence of organic-nitrate (C–ONO₂) and oxygenated phenolic groups in BrC molecular implied the contribution from secondary formation in six megacities, especially in CD city.

Due to the complexity of atmospheric processes, which are usually non-linear in nature, and the traditional linear-based source analytic models may be limited. Here, we used a multilayer perceptron (MLP) model based on artificial neural network (ANN) to improve the source allocation of BrC in these cities. Source apportionment of BrC based on PMF and ANN-MLP analysis revealed that primary emissions (e.g., biomass burning, coal combustion, and vehicle emissions) were key contributors to BrC, and their average contribution in northern cities was about 93.3%, which was 1.2 times higher than that in southern cities. Secondary formation processes made a greater contribution to BrC in southern cities (19.4%) than northern cities (6.7%). The results of our work can provide a basis for the development of more effective practices to control BrC emissions at the regional level.

Data availability. The key data sets are publicly available on the Zendo data repository platform: https://zenodo.org/record/6790321.

Author contribution. ZS: Conceptualization. DW, TZ, and SH: Data curation. ZS and HX: Funding acquisition. DW and YL: Methodology. ZS and JC: Resources. DW: Writing - original draft. DW, ZS, QZ, HX, JS, JC and YL: Writing - review & editing.

Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. This research was supported by the National Natural Science Foundation of China (41877383) and State Key Laboratory of Loess and Quaternary Geology, Institute of Earth Environment, CAS (SKLLQG2103). Authors also thanks for Dr. Jun Tao, Renjian Zhang, Shaofei Kong, and Song Cui for their help in field sampling.

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