Quantifying black carbon light absorption enhancement with a novel statistical approach

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Abstract. Black carbon (BC) particles in the atmosphere can absorb more light when coated by non-absorbing or weakly absorbing materials during atmospheric aging, due to the lensing effect. In this study, the light absorption enhancement factor, $E_{\text{abs}}$, was quantified using a 1-year measurement of mass absorption efficiency (MAE) in the Pearl River Delta region (PRD). A new approach for calculating primary MAE (MAE\textsubscript{p}), the key for $E_{\text{abs}}$ estimation, is demonstrated using the minimum $R^2$ squared (MRS) method, exploring the inherent source independency between BC and its coating materials. A unique feature of $E_{\text{abs}}$ estimation with the MRS approach is its insensitivity to systematic biases in elemental carbon (EC) and $\sigma_{\text{abs}}$ measurements. The annual average $E_{\text{abs}550}$ is found to be 1.50 ± 0.48 (±1 SD) in the PRD region, exhibiting a clear seasonal pattern with higher values in summer and lower in winter. Elevated $E_{\text{abs}}$ in the summertime is likely associated with aged air masses, predominantly of marine origin, along with long-range transport of biomass-burning-influenced air masses from Southeast Asia. Core–shell Mie simulations along with measured $E_{\text{abs}}$ and absorption Ångström exponent (AAE) constraints suggest that in the PRD, the coating materials are unlikely to be dominated by brown carbon and the coating thickness is higher in the rainy season than in the dry season.

1 Introduction

Originating from incomplete combustion, black carbon (BC) is a crucial constituent of atmospheric aerosols, and is an air pollutant itself, having adverse health impacts on humans (Suglia et al., 2008). BC has also been recognized as the third-most important climate forcer due to its broad light absorbing capability across the UV–visible–IR spectrum (IPCC, 2013). BC can alter the climate in a variety of ways, including by direct forcing (Bond et al., 2011), affecting cloud cover (Koch and Del Genio, 2010) and precipitation (Tao et al., 2012), reducing the albedo of snow and ice (Hansen and Nazarenko, 2004) and causing surface dimming (Wild, 2011). The climate effects of BC can be global or regional (Ramanathan and Carmichael, 2008). A recent study found BC can modify planetary boundary layer meteorology, and thus enhance local pollution indirectly (Ding et al., 2016). However, due to its variable optical characteristics induced during atmospheric aging, large uncertainties still exist in estimating the radiative forcing from BC. Optical properties of BC can be predicted by knowing the mass concentration, mixing state and size distribution, which collectively serve as the cornerstone for modeling the climate effect of BC. In 3-D modeling studies, to conserve computational resources, the mass absorption efficiency (MAE) or mass absorption cross section (MAC) are widely used to convert black carbon mass concentration to light absorption co-
efficient ($\sigma_{\text{abs}}$). MAE is a quantity which describes the light absorption ability per unit of elemental carbon (EC) mass:

$$\text{MAE} \left( \text{m}^2 \text{g}^{-1} \right) = \frac{\text{absorption coefficient } \sigma_{\text{abs}} \left( \text{Mm}^{-1} \right)}{\text{EC mass concentration (\(\mu\text{g m}^{-3}\)}}$$  

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As a fundamental input parameter, MAE has a critical impact on the BC radiative forcing estimation in climate modeling studies. Mixing state is one of the governing factors affecting MAE. Light absorption of soot particles is enhanced when coated with non-absorbing materials (Fuller et al., 1999) or weakly absorbing materials (Lack and Cappa, 2010) during atmospheric aging. The coating materials can focus more light onto the soot core through the lensing effect, resulting in elevated MAE (Wang et al., 2017). Strong correlations between MAE and the number/volume fraction of coated particles have been reported in urban areas like Tokyo (Naoe et al., 2009), Shenzhen (Lan et al., 2013) and Xi’an (Wang et al., 2014), implying that the elevated MAE observed at these locations was mainly due to the elevated fraction of coated soot particles. Total absorption ($\sigma_{\text{abs, } t}$) of coated particles can be separated into two parts: primary absorption ($\sigma_{\text{abs, } p}$) due to the uncoated soot core alone, and extra absorption ($\sigma_{\text{abs, } c}$) due to lensing effect of the coating (Bond et al., 2006; Jacobson, 2006; F. Liu et al., 2016) and the presence of secondarily formed brown carbon (BrC; Lack and Cappa, 2010; J. Liu et al., 2016).

$$\sigma_{\text{abs, } t} = \sigma_{\text{abs, } p} + \sigma_{\text{abs, } c}$$

The absorption enhancement factor ($E_{\text{abs}}$) then can be defined as ratio of the total absorption and primary absorption coefficients or the corresponding MAE values:

$$E_{\text{abs}} = \frac{\sigma_{\text{abs, } t}}{\sigma_{\text{abs, } p}} = \frac{\text{MAE}_t}{\text{MAE}_p},$$

where MAE$_p$ represents the ratio of $\sigma_{\text{abs, } p}$ / EC for uncoated soot particles, similar to the concept of the primary OC / EC ratio in the EC tracer method:

$$\text{MAE}_p = \frac{\sigma_{\text{abs, } p}}{\text{EC}}.$$  

and the MAE of coated BC can be defined as follows:

$$\text{MAE}_t = \frac{\sigma_{\text{abs, } t}}{\text{EC}}.$$  

Thus, elevated MAE induced by coating during atmospheric aging results in an $E_{\text{abs}}$ larger than 1.

Previous model studies suggest that absorption by aged soot particles can be 1.5 times greater than fresh soot (Fuller et al., 1999; Bond et al., 2006). Laboratory studies have demonstrated that soot particles coated with secondary organic aerosol (SOA; Saathoff et al., 2003; Schnaiter et al., 2005; Tasoglou et al., 2017) and sulfuric acid (Zhang et al., 2008; Khalizov et al., 2009) can increase $E_{\text{abs}}$. An artificial coating experiment by Shiraiwa et al. (2010) found an $E_{\text{abs}}$ of 2 for graphite particles growing in diameter from 185 to 370 nm. A laboratory study by McMeeking et al. (2014) found that in the presence of BrC, light absorption enhancement is more pronounced at the shorter wavelength. A recent chamber study coupling ambient air with seed BC particles implies that the timescale for $E_{\text{abs}}$, reaching 2.4 is only 5 h in Beijing but 18 h in Houston (Peng et al., 2016). Field studies conducted in recent years have also substantiated enhanced light absorption in Canada (Knox et al., 2009; Chan et al., 2011), USA (Lack et al., 2012b), UK (Liu et al., 2015) and Japan (Nakayama et al., 2014; Ueda et al., 2016). In contrast, field studies in California, USA (Cappa et al., 2012), found a weaker light absorption enhancement (6 % on average). A recent study suggests the mass ratio of non-BC content to BC particles determines the occurrence of the absorption enhancement of black carbon particles (Liu et al., 2017).

Two approaches are widely used to determine $E_{\text{abs}}$ from ambient measurements. The first approach removes the coating materials on particles physically using a thermodenuder (TD; Lack et al., 2012a) or by aerosol filter filtration–dissolution (AFD; X. Cui et al., 2016). The TD approach is briefly discussed here. Coating materials can be removed by TD at a working temperature of around 200 to 300°C (depending on the charring characteristics of aerosols at the sampling site) to measure $\sigma_{\text{abs, } p}$, which is cycled with measurements of $\sigma_{\text{abs, } t}$ (without passing through TD), allowing $E_{\text{abs}}$ to be obtained from the ratio of $\sigma_{\text{abs, } t}/\sigma_{\text{abs, } p}$ following Eq. (3). The major advantage of the TD approach is its ability to provide highly time-resolved measurements (minutes). A photoacoustic spectrometer (PAS) is commonly used with TD for detection to satisfy its high time resolution demands. As an in situ technique, PAS eliminates the artifacts associated with filter-based methods (Weingartner et al., 2003; Coen et al., 2010) and is often considered as the reference instrument for light absorption coefficient determination (Arnott et al., 2003, 2005). One limitation of the TD approach is that a universal optimal operation temperature does not exist, leading to a chemical-composition-dependent efficiency. If the temperature is too low, the coating cannot be fully removed, and charring can occur if the TD temperature is too high, leading to biased results. For example, a soot particle aerosol mass spectrometer (SP-AMS) study in Toronto found that the efficiency of BC coating removal by TD decreased substantially for wildfire-influenced samples (Healy et al., 2015). Another issue is particle loss due to TD, which can be ~20 % and needs to be taken into account (Ueda et al., 2016). It is also worth noting that MAE$_p$ from the TD approach is different from the MAE$_p$ at the emission source. First, the morphology of thermally denuded BC particles (compact aggregates) is different from that of freshly emitted BC particles (chain-like aggregates). Second, most of the coating is removed with the TD denuded soot particles, but freshly emitted soot particles usually come with a thin coating of organic carbon (OC) from condensed
of OC vapors as the temperature drops from engine to the ambient air. As a result, the MAE_\text{p} from the TD approach is expected to be lower than the MAE_\text{p} of the emission source. In this sense, the TD approach may not be a prefect “time machine” to reverse the aging process for \(E_{\text{abs}}\) determination.

The second approach is the MAE ratio method, which is also stated in Eq. (3). The key to this method is determining an appropriate MAE_\text{p} that can represent the MAE from primary emissions. One approach is to adopt the reference MAE_\text{p} from the literature but it may fail to represent the actual MAE_\text{p} at a specific sampling site, since MAE_\text{p} varies temporally and spatially. For example, MAE_\text{p} of diesel soot was found to be 7.1 m^2 g^{-1} at 532 nm (Adler et al., 2010). A much higher MAE_\text{p} (16 m^2 g^{-1} at 530 nm) was observed from natural gas flaring (Weyant et al., 2016). MAE_\text{p} of biomass burning (BB) samples is highly varied due to a wide range of fuel types and combustion conditions (Reid et al., 2005; Roden et al., 2006). A range from 6.1 to 80.8 m^2 g^{-1} was reported for BB MAE_\text{p} at 550 nm (Pandey et al., 2016). Without the knowledge of source contributions, it is not feasible to derive a representative MAE_\text{p} for \(E_{\text{abs}}\) estimation. The other commonly used approach is to determine MAE_\text{p} from the dependency of MAE on the number fraction of coated soot particles measured by SP2 (Lan et al., 2013). Since MAE (y axis) is positively correlated with the number fraction of coated soot particles (x axis), MAE_\text{p} can be determined by extending the regression line to \(x = 0\). It is worth noting that this approach provides only a rough approximation of \(E_{\text{abs}}\) since the parameter used here (coated soot particles number fraction) ignores other main drivers of light absorption enhancement (e.g., coating thickness). As a result, this approach is only valid for a period of measurements for which coating thickness is relatively constant and in which MAE variations are dominated by the coated soot particle number fraction.

However, the high cost of the TD-PAS system and SP2 limits the field measurement of \(E_{\text{abs}}\) around the world. In addition, long-term \(E_{\text{abs}}\) measurements by a TD-PAS system and SP2 are not easily achieved and rarely reported. On the other hand, an Aethalometer and RT-ECOC analyzer can be effectively deployed for long-term measurements and \(E_{\text{abs}}\) estimation, at a relatively lower cost. In this study, based on 1 year of hourly MAE measurements (with the field carbon analyzer and Aethalometer) at a suburban site in the Pearl River Delta (PRD) region of China, quantification of MAE_\text{p} is demonstrated with a novel statistical approach, the minimum R squared method (MRS; Wu and Yu, 2016). The aim of this study is to demonstrate the capability of \(E_{\text{abs}}\) estimation using a 1-year data set with cost-effective instrumentation. The seasonal variability of MAE, absorption Ångström exponent (AAE) and \(E_{\text{abs}}\) in the PRD region are characterized and their dependency on air mass origin and biomass burning are discussed. Abbreviations used in this study are summarized in Table 1 for quick reference.

## 2 Ambient measurements

Sampling was conducted from February 2012 to January 2013 at the suburban Nancun (NC) site (23°0′11.82′′ N, 113°21′18.04′′ E). Nancun, situated at the top of the highest peak (141 m a.s.l.) in Guangzhou’s Panyu district, is located at the geographic center of the PRD region, making it a representative location for average atmospheric mixing characteristics of city clusters located there. Light absorption measurements were performed by a 7λ Aethalometer (AE-31, Magee Scientific Company, Berkeley, CA, USA). The Aethalometer was equipped with a 2.5 µm cyclone with a sampling flow rate of 4 L min^{-1}. The Weingartner algorithm (Weingartner et al., 2003) was adopted to correct the sampling artifacts (aerosol loading, filter matrix and scattering effect) rooted in filter-based method. A customized Aethalometer data processing program (Wu, 2017a) with graphical user interface was developed to perform data correction and detailed descriptions can be found in the Supplement (the program is available from https://sites.google.com/site/wuchengust). Details of the Aethalometer setup and data correction can be found in our previous paper (Wu et al., 2013).

EC mass concentrations were determined by a real-time ECOC analyzer (model RT-4, Sunset Laboratory Inc., Tigard, Oregon, USA). The sunset carbon analyzer sampled at hourly cycles at a flow rate of 8 L min^{-1} with a PM_{2.5} sharp-cut cyclone inlet. For each measurement hour, the first 45 min were for sample collection and the remaining 15 min for thermal–optical analysis. OC is volatized first by stepwise temperature ramping in an oxygen-free atmosphere while in the second stage EC is combusted in the presence of oxygen. Laser transmittance is applied to correct the charing artifact during the OC stage.

Considering a measurement precision of 5 % for the Aethalometer (Hansen, 2005) and 24 % for the RT-ECOC analyzer (Bauer et al., 2009), the propagated relative precision of \(E_{\text{abs}}\) (\(E_{\text{abs}, \text{Unc}}\)) is 35 % following Eqs. (S1) and (S2) in the Supplement. It should be noted that \(E_{\text{abs}, \text{Unc}}\) is mainly attributed to the measurement precision of EC of the RT-ECOC analyzer. Since the measurement precision of the RT-ECOC analyzer estimated by Bauer et al. (2009) is obtained from field measurements at an environment (EC below 1 µg m^{-3}) where EC is much lower than the present study (annual average EC 2.66 ± 2.27 µg m^{-3}), the \(E_{\text{abs}, \text{Unc}}\) of 35 % should be considered as an upper limit for the present study.

Light scattering was measured by an integrating nephelometer (Aurora-1000, Ecotech, Melbourne, Australia). Water-soluble ions were measured by the Monitor for AeRosol and Gases (MARGA; ten Brink et al., 2007). Both instruments are equipped with a PM_{2.5} inlet to remove the coarse particles.
### Table 1. Abbreviations.

<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Definition</th>
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<tr>
<td>AAE&lt;sub&gt;470–660&lt;/sub&gt;</td>
<td>absorption Ångström exponent between 470 and 660 nm</td>
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<tr>
<td>BB</td>
<td>biomass burning</td>
</tr>
<tr>
<td>BrC</td>
<td>brown carbon</td>
</tr>
<tr>
<td>D&lt;sub&gt;core&lt;/sub&gt;, D&lt;sub&gt;shell&lt;/sub&gt;</td>
<td>particle diameter of core/shell</td>
</tr>
<tr>
<td>E&lt;sub&gt;abs,550&lt;/sub&gt;</td>
<td>light absorption enhancement factor at 550 nm</td>
</tr>
<tr>
<td>σ&lt;sub&gt;abs,550&lt;/sub&gt;</td>
<td>light absorption coefficient at 550 nm</td>
</tr>
<tr>
<td>σ&lt;sub&gt;abs,t&lt;/sub&gt;</td>
<td>total light absorption coefficient of a coated particle</td>
</tr>
<tr>
<td>σ&lt;sub&gt;abs,p&lt;/sub&gt;</td>
<td>primary light absorption coefficient attributed to the soot core alone of a coated particle</td>
</tr>
<tr>
<td>σ&lt;sub&gt;abs,c&lt;/sub&gt;</td>
<td>extra light absorption other than σ&lt;sub&gt;abs,p&lt;/sub&gt;, including those from the lensing effect of coating on the soot core and brown carbon</td>
</tr>
<tr>
<td>LII</td>
<td>laser-induced incandescence technique for soot measurement</td>
</tr>
<tr>
<td>MAE&lt;sub&gt;550&lt;/sub&gt;</td>
<td>mass absorption efficiency at 550 nm, also known as mass absorption cross section (MAC)</td>
</tr>
<tr>
<td>MAE&lt;sub&gt;p,550&lt;/sub&gt;</td>
<td>primary MAE of freshly emitted soot particles at 550 nm</td>
</tr>
<tr>
<td>MAAP</td>
<td>multi-angle absorption photometer</td>
</tr>
<tr>
<td>MOUDI</td>
<td>micro-orifice uniform deposit impactor</td>
</tr>
<tr>
<td>MRS</td>
<td>minimum R squared method</td>
</tr>
<tr>
<td>PAS</td>
<td>photoacoustic spectrometer</td>
</tr>
<tr>
<td>PRD</td>
<td>Pearl River Delta region, China</td>
</tr>
<tr>
<td>SP2</td>
<td>single-particle soot photometer</td>
</tr>
<tr>
<td>SSA</td>
<td>single-scattering albedo</td>
</tr>
<tr>
<td>TD</td>
<td>thermodenuder</td>
</tr>
<tr>
<td>TOA</td>
<td>thermal–optical analysis</td>
</tr>
<tr>
<td>TSV</td>
<td>total spatial variance in backward trajectories cluster analysis</td>
</tr>
</tbody>
</table>

#### 2.1 Uncertainties of MAE determination

Two major uncertainties associated with the σ<sub>abs</sub> and EC determination techniques should be taken into account when comparing MAE across different studies. For the σ<sub>abs</sub> determination technique, photoacoustic spectroscopy (PAS) is an in situ technique free from filter-based artifacts, but its application is limited by its high cost. The filter-based optical transmittance method (e.g., Aethalometer and multi-angle absorption photometer, MAAP) is the most widely used technique around the world, but data correction is needed to minimize the bias from artifacts due to the loading, matrix and scattering effects (Weingartner et al., 2003; Arnott et al., 2005; Schmid et al., 2006; Virkkula et al., 2007; Coen et al., 2010; Drinovec et al., 2017; Saturno et al., 2017). Besides these artifacts, RH is also a source of σ<sub>abs</sub> measurement uncertainty. Elevated RH is not only a driving force of increased σ<sub>abs</sub> due to the hygroscopic growth of particles but also a factor affecting ambient σ<sub>abs</sub> measurements. Previous studies found σ<sub>abs</sub> from PAS exhibit a systematic decrease when RH > 70% (Arnott et al., 2003; Kozlov et al., 2011). Water evaporation was found as the major cause for the biased PAS σ<sub>abs</sub> measurements under high RH (Raspet et al., 2003; Lewis et al., 2009b; Langridge et al., 2013). Filter-based measurements are also affected under high RH conditions. For example, Arnott et al. (2003) observed erratic responses from a particle soot absorption photometer (PSAP) as RH changed. The main reason is traced to the hydrophilic cellulose membrane, which serves to reinforce the quartz filter used in the PSAP. The fibers can swell and shrink as RH changes, causing unwanted light attenuation signal. The PTFE-coated fiberglass tape has been available since 2012 for recent model of Aethalometers to minimize RH interference (Magee Scientific, 2017). A study by Schmid et al. (2006) reported dependency of the PSAP σ<sub>abs</sub> on RH, but found negligible effect of RH on Aethalometer performance. It is also worth noting that RH in the Aethalometer optical chamber may be lower than the ambient RH due to the slightly elevated temperature inside the instrument. The magnitude of RH difference was found to be similar between different instruments: 20% for the Aethalometer (Schmid et al., 2006) and 15% for the nephelometer (Guyon et al., 2004). The RH in the Aethalometer optical chamber was not measured in this study. We expected its level to be slightly lower than the ambient RH. Cappa et al. (2008) found σ<sub>abs</sub> measurements from PSAP and PAS maintained a high linearity even under high RH conditions (65–91%). Inter-comparison studies demonstrated that with proper corrections, Aethalometer σ<sub>abs</sub> measurements agree well with those from PAS (Ajtai et al., 2011). During the inter-comparison study of an Aethalometer (AE-16) and a PAS in Guangzhou (Wu et al., 2009), good correlation was found (R² = 0.96) as shown in Fig. S1 in the Supplement. These comparison results imply that the Aethalometer results are linearly correlated with PAS measurements and RH has a limited interference on Aethalometer measurements. In our study, careful corrective measures...
(Wu et al., 2013) are conducted for the Aethalometer \( \sigma_{\text{abs}} \) data treatment to minimize these artifacts, but they still cannot be fully eliminated.

For the EC determination, different thermal–optical analysis (TOA) protocols can impact the measurement variability and thus MAE. As shown in Table S1, MAE for the same samples at Fresno varied from 6.1 to 9.3 m\(^2\) g\(^{-1}\), depending on which EC analysis protocol was applied (Chow et al., 2009). Studies in the PRD found that discrepancies in EC measured by different analysis protocols could be as large as a factor of 5 (C. Wu et al., 2012, 2016), which adds to the uncertainty for the MAE estimation. In addition, EC from TOA is also different from refractory BC (rBC) reported by the laser-induced incandescence (LII) technique (e.g., single-particle soot photometer, SP2). For example, two studies in Toronto (Knox et al., 2009; Chan et al., 2011) both used the PAS for \( \sigma_{\text{abs}} \) measurement but different techniques for EC mass determination, resulting in very different MAE results. LII instruments are usually calibrated with a commercially available surrogate (e.g., fullerene) since direct calibration with ambient soot is not easy to achieve. Laborde et al. (2012) indicates that the incandescence response of SP2 exhibits a dependency on soot type (15% between fullerene and denuded diesel soot particles; 14% between biomass burning and denuded diesel soot particles). Due to the absence of widely accepted reference materials for EC, the uncertainties in EC determination will exist in the foreseeable future. All these uncertainties, including the uncertainty of rBC mass determination from SP2, uncertainty of EC in TOA, the discrepancy between SP2 rBC and TOA EC and the discrepancy of \( \sigma_{\text{abs}} \) between filter transmission and photoacoustic methods, can contribute to the differences in MAE listed in Table S1.

Systematic bias in MAE (e.g., overestimation of \( \sigma_{\text{abs}} \) and variability of EC mass with different TOA protocols) discussed above have little effect on \( E_{\text{abs}} \) estimation with MRS. As shown in Eq. (3), \( E_{\text{abs}} \) is the ratio of MAE\(_t\) to MAE\(_p\) or \( \sigma_{\text{abs}, t} \) to \( \sigma_{\text{abs}, p} \); thus most of the bias in EC mass or \( \sigma_{\text{abs}} \) is canceled out during the \( E_{\text{abs}} \) calculation. More details are discussed in Sect. 5.1.

3 Methodology

3.1 MAE\(_p\) estimation with MRS from the ambient data

In this section, a new approach for MAE\(_p\) estimation is introduced for \( E_{\text{abs}} \) determination, which requires knowledge of differentiating \( \sigma_{\text{abs}, p} \) and \( \sigma_{\text{abs}, c} \) portions in \( \sigma_{\text{abs}, t} \). The idea of decoupling \( \sigma_{\text{abs}, t} \) into \( \sigma_{\text{abs}, p} \) and \( \sigma_{\text{abs}, c} \) is conceptually similar to decoupling OC into primary OC (POC) and secondary OC (SOC) in the EC tracer method as shown in Table 2. In the EC tracer method, if (OC/EC)\(_p\) is known, POC can be determined from OC (Turpin and Huntzicker, 1991). The role of MAE\(_p\) here is similar to the role of (OC/EC)\(_p\), the primary OC/EC ratio in the EC tracer method (a comparison is given in Table 2). If MAE\(_p\) (average MAE from primary emission sources) is known, \( E_{\text{abs}} \) can be obtained from the ratio of MAE\(_t\)/MAE\(_p\) (Eq. 3). Therefore, the key for \( E_{\text{abs}} \) estimation is to derive an appropriate MAE\(_p\). It is worth noting that MAE\(_p\) here does not represent MAE from a single or specific primary emission source; instead, it reflects an average and effective MAE that has taken consideration of various primary emission sources. Thus, the MAE\(_p\) is conceptually analogous to (OC/EC)\(_p\) in the EC tracer method in which the primary ratio reflects an overall ratio from primary emission sources rather than from a single primary source.

The minimum \( R \) squared method (MRS) explores the inherent independency between pollutants from primary emissions (e.g., EC) and products associated with secondary formation processes (e.g., SOC, \( \sigma_{\text{abs}, c} \)) to derive the primary ratios (e.g., (OC/EC)\(_p\), MAE\(_p\)) in the EC tracer method (Wu and Yu, 2016). When applying MRS for light absorption enhancement estimation, MRS is used to explore the inherent independency between EC and \( \sigma_{\text{abs}, c} \), which is gained during atmospheric aging after emission. An example of MAE\(_p\) estimation with MRS is shown in Fig. 1. Firstly, the assumed MAE\(_p\) value is varied continuously over a reasonable range (0.01 to 50 m\(^2\) g\(^{-1}\) as shown in Fig. 1). Then at each hypothetical MAE\(_p\), \( \sigma_{\text{abs}, c} \) can be calculated by Eq. (6) (a combination of Eqs. 2 and 4) using EC and \( \sigma_{\text{abs}, t} \) from ambient measurements.

\[
\sigma_{\text{abs}, c} = \sigma_{\text{abs}, t} - \text{MAE}_{p} \times \text{EC}
\]

Accordingly, for each hypothetical MAE\(_p\), a correlation coefficient value (\( R^2 \)) of \( \sigma_{\text{abs}, c} \) vs. EC (i.e., \( R^2(\sigma_{\text{abs}, c}, \text{EC}) \)) can be obtained. The series of \( R^2(\sigma_{\text{abs}, c}, \text{EC}) \) values (y axis) are then plotted against the assumed MAE\(_p\) values (x axis) as shown by the red curve in Fig. 1. The physical meaning of this plot can be interpreted as follows. The \( \sigma_{\text{abs}, p} \) is the fraction of light absorption due to primary emitted soot particles. As a result, \( \sigma_{\text{abs}, p} \) is well correlated with EC mass. In contrast, the \( \sigma_{\text{abs}, c} \) is the fraction of light absorption gained by the lensing effect of the coating on particles after emission. The variability of \( \sigma_{\text{abs}, c} \) mainly depends on the coating thickness of the soot particles. Consequently, \( \sigma_{\text{abs}, c} \) is independent of EC mass. Since variations of EC and \( \sigma_{\text{abs}, c} \) are independent, the assumed MAE\(_p\) corresponding to the minimum \( R^2 \) (EC, \( \sigma_{\text{abs}, c} \)) would then represent the most statistically probable MAE\(_p\) of the tested data set.

A computer program (Wu, 2017b) in Igor Pro (WaveMetrics, Inc. Lake Oswego, OR, USA) was developed to facilitate MRS calculation with a user friendly graphical user interface. Another two Igor Pro-based computer programs, Histbox (Wu, 2017c) and Scatter Plot (Wu, 2017d), are used for generating the histograms, box plots and scatter plots (with Deming regressions) presented in this study. Detailed descriptions of these computer programs can be found in the Supplement and the computer programs are available from https://sites.google.com/site/wuchengust.
Table 2. Comparison of MRS application on (OC / EC)_p (for SOC estimation) and MAE_p (for E_{abs} estimation).

<table>
<thead>
<tr>
<th>Key parameter of fresh EC particles to be determined</th>
<th>Input quantities for MRS from measurements</th>
<th>Variable to be decoupled by the tracer</th>
<th>Ambient measurement at its closest to fresh emissions</th>
<th>Graph</th>
</tr>
</thead>
<tbody>
<tr>
<td>MRS in EC tracer method for SOC estimation (Wu and Yu, 2016)</td>
<td>(OC / EC)_p = \frac{POC}{EC}</td>
<td>OC, EC (tracer)</td>
<td>SOC = OC - \left(\frac{OC}{EC}\right)_p \times EC</td>
<td>Minimum R^2 (SOC, EC)</td>
</tr>
<tr>
<td>MRS in EC tracer method for E_{abs} estimation (this study)</td>
<td>\sigma_{abs, p, EC}</td>
<td>\sigma_{abs, t, EC} (tracer)</td>
<td>\sigma_{abs, t} = \sigma_{abs, p} + \sigma_{abs, c}</td>
<td>Minimum R^2 (\sigma_{abs, c, EC})</td>
</tr>
</tbody>
</table>

\[
\sigma_{abs, c} = \frac{\sigma_{abs, t}}{\sigma_{EC}} \times EC + \sigma_{abs, c}
\]

\[
R^2 (SOC/EC)_{Min} = 2.26
\]

\[
R^2 (\sigma_{abs, c, EC})_{Min} = 13
\]

![Figure 1. Minimum R squared (MRS) plot for calculating MAE_p at 550 nm. The red curve is the correlation result between \(\sigma_{abs, c} (\sigma_{abs, t} - EC \times MAE_p)\) and EC mass. The shaded area in light tan represents the frequency distribution of observed MAE. The dashed green line is the cumulative distribution of observed MAE.](image)

3.2 Mie simulation

It can be informative to model a single soot particle using Mie theory (Bohren and Huffman, 1983) and understand the theoretical range and variability of the soot particle’s optical properties. Three mixing states are widely employed for parameterization: internal, external and core–shell mixing. To better represent reality (coating due to the aging process), a core–shell model is considered in the Mie calculation (Fig. S2), which is more realistic than a volume mixture model (Bond et al., 2006). An aerosol optical closure study in the North China Plain (NCP) found that the core–shell model can provide better performance than assuming purely internal mixing and external mixing (Ma et al., 2012). A morphology study using Scanning transmission X-ray microscopy found that core–shell is the dominating mixing state in ambient samples (Moffet et al., 2016). It should be noted that the core–shell model assumption still has its own limitations. A single-particle soot photometer (SP2) study by Sedlacek et al. (2012) reported a negative lag time between the scattering and incandescence signals in samples influenced by biomass burning, implying a near-surface location of soot relative to non-absorbing materials. Near-surface mixing of soot has also been observed in Tokyo, but accounted for only 10% of total mixed soot containing particles (Moteki et al., 2014). Considering the domination of core–shell-type particles in the ambient environment, the core–shell assumption in our optical model is sufficient to approximate reality.

As shown in Fig. S2, freshly emitted soot particles are chain-like aggregates of small spheres (∼30–50 nm). After the aging process, soot particles are coated with organic and inorganic materials. Sufficient evidence has shown that the coating not only results in particle size growth but also makes the soot core become more compact due to its collapse (Alexander et al., 2008; Zhang et al., 2008; Lewis et al., 2009a), especially under high RH conditions (Leung et al., 2017). A recent study by Pei et al. (2017) shown that fill-
ing of void space within the agglomerate is the first step of the morphological transformation of soot particles in atmospheric aging, leading to a spherical soot core. Since the spherical core and shell favor Mie simulation, both core and shell are considered as spheres in the Mie calculation.

To investigate the spectrum properties of soot particles, 11 wavelengths (370, 405, 470, 520, 532, 550, 590, 660, 781, 880 and 950 nm) are considered in calculations to cover wavelengths in the most frequently used absorption measurement instruments. A refractive index (RI) of 1.85–0.71 is adopted for soot core (Bond and Bergstrom, 2006) and 1.55 for non-absorbing coating (clear shell) in the Mie calculation for all wavelengths. Studies suggest a group of organic matter (OM), known as brown carbon (BrC), can absorb solar radiation at UV wavelengths (Kirchstetter et al., 2004). Thus, a BrC coating (brown shell) scenario is also considered in Mie simulations following the wavelength-dependent RI suggested by Lack and Cappa (2010), which ranges from 1.55–0.059i (370 nm) to 1.55–0.0005i (950 nm).

A modeling study by Bond et al. (2006) indicates that absorption amplification is not sensitive to the RI; thus the result below is not expected to be sensitive to the RI variability. Due to the spherical assumption of the BC core, a constant particle density is adopted for simplicity instead of size-dependent particle density. But it is worth noting that in reality, the effective density of soot varies with particle size due to the morphology change during particle aging (Tavakoli and Olfter, 2014; Dastanpour et al., 2017). Both core diameters ($D_{\text{core}}$) and shell diameters ($D_{\text{shell}}$) are constrained in the range of $\sim$10–3000 nm in the model simulations. The Mie calculations are implemented with a customized program (Wu, 2017e) written in Igor Pro (WaveMetrics, Inc. Lake Oswego, OR, USA) and it is available from https://sites.google.com/site/wuchengust. It should be noted that the core–shell–type mixing state of particles is still rare in 3-D atmospheric models like WRF-Chem (Matsui et al., 2013; Nordmann et al., 2014) due to computational cost limitation.

### 3.2.1 Mie-modeled absorption Ångström exponent (AAE)

The absorption Ångström exponent (AAE) is a widely used parameter that describes the wavelength dependence of aerosol light absorption (Moosmuller et al., 2011), which can be written explicitly as

$$\text{AAE}(\lambda_1, \lambda_2) = -\frac{\ln (\sigma_{\text{abs}, \lambda_1}) - \ln (\sigma_{\text{abs}, \lambda_2})}{\ln (\lambda_1) - \ln (\lambda_2)}. \quad (7)$$

It is well known that ambient soot particles exhibit an AAE close to unity (Bond, 2001). Modeled variability in AAE$_{470–660}$ of bare soot particles is shown in Fig. S3. For soot particles with $D_{\text{core}} < 200$ nm, AAE$_{470–660}$ is very close to 1 and increases significantly for particles with $D_{\text{core}} > 200$ nm. Considering a typical $D_{\text{core}}$ of freshly emitted soot particles smaller than 200 nm (Rose et al., 2006; China et al., 2013), the model results confirm the frequently observed AAE close to 1 from ambient measurements (Kirchstetter et al., 2004). Modeled variability in AAE$_{470–660}$ of soot particles coated by non-absorbing substances (clear shell) and weakly absorbing materials (brown shell) is shown in Fig. 2. An absorption Ångström exponent value elevated to $\sim 2$ is observed in the clear shell scenario (Fig. 2a and b) for the most probable soot core particle sizes ($< 200$ nm), which agrees well with a previous model study (Lack and Cappa, 2010), implying that elevated AAE cannot be exclusively attributed to mixing with BrC. AAE elevation is more pronounced in the brown shell scenario. For soot particles with $D_{\text{core}} < 200$ nm, brown shell AAE$_{470–660}$ can easily reach 3 for a coating of $D_{\text{shell}}/D_{\text{core}} = 3$ (Fig. 2c and d). These high AAE results are consistent with the previous model study (Lack and Cappa, 2010) and could partially explain the high AAE observed in measurement studies (Kirchstetter et al., 2004; Hoffer et al., 2006), since the presence of externally mixed BrC particles also contribute to the wavelength-dependent light absorption.

### 3.2.2 Mie-modeled single-scattering albedo (SSA)

Variability in modeled SSA$_{525}$ of soot particles coated by non-absorbing substances and weakly absorbing materials (e.g., BrC) is shown in Fig. S4. For particles with $D_{\text{core}} < 200$ nm and $D_{\text{shell}}/D_{\text{core}} < 3$, the SSA increases gradually (up to $\sim 0.9$) with a thicker coating and behaves similarly in clear shell and brown shell scenarios.

### 3.2.3 Mie-modeled mass absorption efficiency (MAE)

MAE is a useful indicator for soot mixing state. Variability in MAE of bare soot particles as a function of particle size at a wavelength of 550 nm is illustrated in Fig. S5. The magnitude of MAE is sensitive to the soot density assumption, especially for particles $< 200$ nm (Fig. S5), but the overall trend of particle size dependency is similar among different density scenarios. MAE peaks at a particle size of 200 nm and decreases dramatically for larger particles. In our MAE calculation, a soot density of 1.9 g cm$^{-3}$ is adopted, as suggested by Bond and Bergstrom (2006). The purpose of adopting constant density is to simplify the MAE calculation. It should be noted that the effective density of soot core is highly variable in ambient environments. For example, a study in Beijing (Y. Zhang et al., 2016) found a value of 1.2 g cm$^{-3}$. A recent chamber study found the effective density of soot can evolve from 0.43 to 1.45 g cm$^{-3}$ during aging as coated by $m$-Xylene oxidation products (Guo et al., 2016).

A study using a single-particle aerosol mass spectrometer in Guangzhou found the effective density of soot increased with particle size in the range of 400 to 1600 nm (G. Zhang et al., 2016). The MAE of coated particles from different core/shell diameter combinations are shown in Fig. S6. For
thickly coated particles, the MAE in the clear shell scenario varied as $D_{\text{shell}}/D_{\text{core}}$ increased, but the MAE of brown shell scenario increased quasi-monotonously with $D_{\text{shell}}/D_{\text{core}}$.

3.2.4 Mie-modeled light absorption enhancement factor ($E_{\text{abs}}$)

$E_{\text{abs}}$ is a better indicator for soot mixing state than MAE since it does not rely on the soot density assumption and is more suitable for comparison of Mie simulations with ambient measurements. Modeled variability in $E_{\text{abs}}$ of soot particles coated by non-absorbing substances and weakly absorbing materials (e.g., BrC) is shown in Fig. 3a and c, respectively. $E_{\text{abs}}$ is not only sensitive to the core/shell diameter combination, but it also behaves very differently with the clear and brown shell assumptions. For the clear shell scenario, when $D_{\text{coat}}/D_{\text{core}} < 2$, $E_{\text{abs}}$ does not exceed 2 for particles with different soot core sizes, but with the same $D_{\text{coat}}/D_{\text{core}}$, a larger soot core size yields a higher $E_{\text{abs}}$ (Fig. 3b, cross sections of Fig. 3a). If $D_{\text{coat}}/D_{\text{core}} > 2$, $E_{\text{abs}}$ could be 3 to 5 for particles with $D_{\text{core}} > 200$ nm, but for particles with a soot core larger than 200 nm, the $E_{\text{abs}}$ would be limited to $\sim 2$ as shown in Fig. 3b. For the brown shell scenario, $E_{\text{abs}}$ increased quasi-monotonically with $D_{\text{coat}}/D_{\text{core}}$, and this trend is similar for different soot core sizes (Fig. 3d).

The main reason is that in the brown shell scenario, both lensing effect and BrC absorption contribute to $E_{\text{abs}}$. As shown in Fig. S7, the BrC absorption contribution to total $E_{\text{abs}}$ strongly depends on coating thickness and is insensitive to soot core diameters. When the coating is relatively thin ($< 5$ nm for $\lambda$ at 370 nm, $< 15$ nm for $\lambda$ at 550 nm and $< 40$ nm for $\lambda$ at 880 nm), BrC absorption contribution to the total $E_{\text{abs}}$ is less than 20%. As the coating increases to a certain level ($\sim 15$ nm for $\lambda$ at 370 nm, $\sim 35$ nm for $\lambda$ at 550 nm and $\sim 90$ nm for $\lambda$ at 880 nm), BrC absorption contribution.

Figure 2. Mie simulated size dependency of AAE$_{470-660}$ soot particles. (a) Combination of different clear shell (y axis) and core diameters (x axis). The color coding represents the AAE$_{470-660}$ of a particle with specific core and clear shell size; (b) cross-section views of (a). The color coding represents different $D_{\text{core}}$ in the range of $\sim 50$–300 nm. (c, d) Similar to panels (a) and (b) but for the brown shell scenario.
Figure 3. Mie simulated size dependency of soot particles $E_{abs}$ at wavelength 550 nm. (a) Combination of different clear shell ($y$ axis) and core diameters ($x$ axis). The color coding represents the $E_{abs}$ of a particle with specific core and clear shell size; (b) cross-section views of (a). The color coding represents different $D_{core}$ in the range of 50–300 nm. (c, d) Similar to panels (a) and (b) but for the brown shell scenario.

is comparable to the lensing effect contribution, each contributing $\sim 50\%$ to the total $E_{abs}$. When the BrC coating is sufficiently thick ($> 30$ nm for $\lambda$ at 370 nm, $> 90$ nm for $\lambda$ at 550 nm and $> 110$ nm for $\lambda$ at 880 nm), BrC absorption dominates the $E_{abs}$ contribution. As a result, if BrC coating is indeed present in ambient samples, a strong wavelength-dependent $E_{abs}$ can be observed, since a BrC coating of 30 nm is enough to induce a large amount of detectable $E_{abs}$ in the UV range. Another major difference between the clear and brown shell scenarios is that, for thickly coated particles (e.g., $D_{coat}/D_{core} > 2$), the brown shell can yield a much higher $E_{abs}$ than the clear shell.

Both primary soot size distribution and coating thickness can affect the absorption enhancement of ambient BC particles. Ambient measurements from LII found soot particle number and mass modes peaking at 110 and 220 nm, respectively, in the PRD (Huang et al., 2011). A study in Shanghai found similar results (70 nm for number concentrations and 200 nm for mass concentrations; Gong et al., 2016). Considering that the LII technique is specific to BC mass determination which is independent of BC mixing state, the size distribution reported using LII is able to represent the size distribution of the BC core. A study using a micro-orifice uniform deposit impactor (MOUDI) found a EC mass size distribution in the PRD exhibiting three modes peaking at $\sim 300$, $\sim 900$ and $\sim 5000$ nm (Yu et al., 2010), implying a substantial coating of BC particles, and a diameter amplification of 3. BC sizing from LII is based on volume-equivalent diameter (VED), while MOUDI is based on aerodynamic diameter. As a result, these two techniques do not necessarily yield similar sizes, even for the bare soot particles. The conversion between these two types of diameters involves the knowledge of particle density and morphology (drag force). A recent closure study on BC mixing state in the PRD region sug-
suggests $\sigma_{abs}$ is dominated by coated soot particles in the range of $\sim$300–400 nm (Tan et al., 2016). Considering the dominant BC core distribution measured with SP2 (110 nm), the upper limit of $E_{abs}$ in the PRD is roughly estimated as $\sim 2$ for the clear shell scenario (Fig. 3b).

4 Results and discussions

4.1 Annual measurement statistics

The frequency distribution (lognormal) of $\sigma_{abs550}$ is shown in Fig. 4a, with an annual average ($\pm 1$ SD) of 42.65 ± 30.78 Mm$^{-1}$. A lognormal distribution is also found in the EC mass concentration (Fig. 4b), with an annual average of 2.66 ± 2.27 μg m$^{-3}$. Figure 4c demonstrates the 1-year frequency distribution of MAE$_{550}$ at the NC site. The annual average MAE$_{550}$ is 18.75 ± 6.16 m$^2$ g$^{-1}$ and the peak ($\pm 1$ SD) of the lognormal fit is 15.70 ± 0.22 m$^2$ g$^{-1}$. A good correlation is observed between $\sigma_{abs}$ and EC mass ($R^2 = 0.92$) as shown in Fig. 4d, and the color coding indicates a MAE dependency on RH, which agrees with a study in Xi’an (Y. Wu et al., 2016). Annual average AAE$_{525}$ is 0.86 ± 0.05 (Fig. S8c), similar to previous studies in the PRD (Jung et al., 2009; Wu et al., 2009). For comparison purpose, MAE measured at original wavelength and MAE scaled to 550 nm following the $\lambda^{-1}$ assumption are both shown in Table S1. The MAE comparisons discussed below are MAE at 550 nm. MAE$_{550}$ in previous studies at various locations was found to cover a wide range, from 5.9 to 61.6 m$^2$ g$^{-1}$. Annual average observed MAE$_{550}$ at NC (18.75 m$^2$ g$^{-1}$) is higher than many studies shown in Fig. 5, e.g., Shenzhen (Lan et al., 2013), Beijing (Yang et al., 2009), Mexico City (Doran et al., 2007) and Fresno (Chow et al., 2009).

As shown in Fig. 1, the annual average MAE$_{p,550}$ estimated from MRS is 13 m$^2$ g$^{-1}$. MAE$_p$ from MRS represents the MAE at the emission source, which is different from the MAE$_p$ found in the TD approach for two reasons. First, the morphology of thermally denuded BC particles (compact aggregates) is different from that of freshly emitted BC particles (chain-like aggregates). Second, most of the coatings are removed for TD denuded soot particles, but freshly emitted soot particles usually come with a thin coating of OC formed from condensation of OC vapors as the temperature drops from the engine to the ambient air. As a result, the MRS-derived MAE$_p$ is expected to be higher than the MAE$_p$ in the

Figure 4. Measured annual statistics of $\sigma_{abs550}$, EC and MAE$_{550}$. (a) Annual frequency distribution of light absorption at 550 nm. The red curve represents the fitting line for a lognormal distribution. (b) Annual frequency distribution of EC mass concentration. (c) Frequency distribution of Mass absorption efficiency (MAE) at 550 nm. (d) Scatter plot of light absorption (550 nm) and EC mass. The slope represents MAE$_{550}$. The blue regression line is made using Deming regression. The color coding represents RH.
TD approach. The estimated MAE_p,550 is higher than a previous study in Guangzhou (7.44 m^2 g^{-1}; Andreae et al., 2008), but comparable to Xi’an (11.34 m^2 g^{-1}; Wang et al., 2014) and Toronto (~9.53–12.57 m^2 g^{-1}; Knox et al., 2009). The annual average E_{abs,550} from MRS following Eq. (3) is estimated to be 1.50 ± 0.48 (mean ±1 SD).

As mentioned in Sect. 1, the definition of MAE_p from the TD approach is different from the MAE_p of the emission source. The TD MAE_p is expected to be slightly lower than the MAE_p of the emission source. Therefore, the corresponding E_{abs} values are slightly different and care should be taken when comparing MRS-derived E_{abs} with E_{abs} from the TD approach and Mie simulations. The E_{abs} could vary by location, depending on the coating thickness and size distribution of the primary aerosols. After undergoing atmospheric aging, the E_{abs} can increase during transport from emission source to rural areas. The magnitude of the E_{abs} found at the NC site is comparable to other locations such as Boulder (Lack et al., 2012a; 1.38), London (Liu et al., 2015; 1.4), Shenzhen (Lan et al., 2013; 1.3), Yuncheng (X. Cui et al., 2016; 2.25), Jinan (Chen et al., 2017; 2.07) and Nanjing (F. Cui et al., 2016; 1.6) and is higher than studies in California (Cappa et al., 2012; 1.06), as listed in Table 3. Spectrum E_{abs} values are calculated from 370 to 950 nm as shown in Fig. S9. E_{abs} in the PRD exhibits a weak wavelength dependence, with slightly higher E_{abs} at the shorter wavelength (e.g., E_{abs,370} = 1.55 ± 0.48) and relatively lower in the IR range (e.g., E_{abs,950} = 1.49 ± 0.49).

4.2 Monthly characteristics of MAE, AAE and SSA

Monthly variations of MAE_{550} at the NC site are shown in Fig. 6a and Table S2, revealing distinct patterns of higher MAE_{550} in summer and lower in winter. On the other hand, AAE_{470–660} is lower in summer and higher in winter (Fig. 6b and Table S3). Monthly SSA_{525} varied from 0.83 to 0.90 without a clear seasonal pattern, as shown in Fig. S10 and Ta-
4.3 The effect of air mass origin

Understanding the seasonal variations of optical properties in the PRD is of great interest. Hourly backward trajectories for the past 72 h were calculated using NOAA’s HYSPLIT (Hybrid Single Particle Lagrangian Integrated Trajectory, version 4) model (Draxier and Hess, 1998) from February 2012 to January 2013 as shown in Fig. S11. Cluster analysis was conducted using MeteoInfo (Wang, 2014). By examining the total spatial variance (TSV), the number of clusters was determined to be four as shown in Fig. S12. Cluster 1 (C1) represents continental air masses from the north, accounting for 44.4 % of total trajectories. C2 (22.8 %) represents marine air masses coming from the South China Sea. C3 represents air masses coming from the east (Taiwan), C4 (15.8 %) represents transitional air masses coming from the east coastline of China. As shown in Fig. 7, $E_{abs}^{550}$ from C2 (1.78) is higher than other clusters (1.30–1.42). Further Wilcoxon–Mann–Whitney tests show that $E_{abs}^{550}$ from C2 is significantly higher than $E_{abs}^{550}$ from C1, C3 and C4 (Fig. S13), implying that particles from the South China Sea cluster is likely more aged than other clusters. Air mass origin in the PRD is dominated by C2 from April to August (Fig. S14a) as a result of the South China Sea monsoon in the rainy season. In contrast, the dry season is ruled by continental air masses from the north (C1) due to the influence of the northeast monsoon. $E_{abs}^{550}$ from C2 varied from 1.67 to 2.19, but was always higher than $E_{abs}^{550}$ from C1 and C3 during the rainy season (Fig. S14b). As a result, the domination of aged air mass from the vast ocean is one of the reasons for the much higher $E_{abs}^{550}$ found in the rainy season.

4.4 The effect of biomass burning

Biomass burning (BB) and vehicular emission are the two major sources of soot particles. BC from biomass burning emission, depending on the fuel type and burning condition, may have a higher OC / EC ratio and a thicker coating, resulting in a higher MAE than vehicular emission (Shen et al., 2013; Cheng et al., 2016). In this study, the influence of BB on optical properties is investigated using the $K^{+}$ / EC ratio as a BB indicator. As shown in Fig. 8, MAE$_{550}$ is positively correlated with the $K^{+}$ / EC ratio, which exhibits a clear seasonal pattern that is higher in the rainy season and lower in the dry season (Fig. S15a). Southeast Asia has the highest fire emission density globally due to the high biofuel consumption along with frequent fire activity in this region (Aouizerats et al., 2015), making Southeast Asia a large contributor to BC emissions (Jason Blake, 2014). During the rainy season when oceanic wind prevails, BC from BB emissions in Southeast Asia can reach PRD through long-range transport, resulting in an elevated $K^{+}$ / EC ratio and MAE$_{550}$. The Deming regression intercept (11.89) in Fig. 8 represents the MAE without the BB effect. This non-BB MAE$_{550}$ (11.89 m$_{2}$ g$^{-1}$) is only slightly lower than MAE$_{p,550}$ (13 m$_{2}$ g$^{-1}$) obtained in Sect. 4.1, implying that a large frac-
tion of MAE$_{p,550}$ could not be explained by the BB source. Additional evidence was obtained through examining regression relationships of MAE$_{p,550}$ with K$^+$/EC month by month (Fig. S15b). Correlation of monthly MAE$_{p,550}$ vs. K$^+$/EC ratio yields an $R^2$ of 0.23 (Fig. S15c). In contrast, a much higher correlation ($R^2 = 0.58$) was observed (Fig. S15d) between MAE$_{p,550}$ and non-BB MAE$_{550}$ (i.e., K$^+$/EC intercepts from Fig. S15b). These results imply that BB is one of the contributors to the MAE$_{p,550}$ variations, but unlikely the dominating one.

Many studies have found that BB-influenced samples exhibit elevated AAE due to the presence of wavelength-dependent light absorbing substances like BrC and humic-like substances (HULIS; Kirchstetter et al., 2004; Hoffer et al., 2006; Sandradewi et al., 2008; Herich et al., 2011; Pokhrel et al., 2017). Investigating whether elevated AAE observed in the PRD during the dry season is associated with BB influence is of great interest. As shown in Fig. S16, AAE$_{470-470}$ and AAE$_{470-660}$ did not correlate with the BB indicator, K$^+$/EC ratio. These results suggest that the elevated AAE observed in the PRD wintertime is unlikely to be dominated by the BB effect. Beside the independence between AAE$_{470-660}$ and K$^+$/EC ratio, the measured AAE$_{470-660}$ range also implies that BB is not the major driving force of AAE$_{470-660}$ variations. The limited light absorption contribution from BrC in PRD region is observed in a recent study (Yuan et al., 2016), which suggest an upper limit of BrC contribution of 10% at 405 nm in the winter time using the AAE approach. As discussed in our Mie simulation (Sect. 3.2) and a previous study (Lack and Cappa, 2010), coating of non-absorbing materials onto soot particles can increase AAE up to 2. Since the monthly average AAE$_{470-660}$ in wintertime did not exceed 1.2 (Table S3), the variations of AAE$_{470-660}$ in the PRD are more likely associated with coatings rather than the contribution of BrC. The results also imply that attempts at attributing BrC absorption for the PRD data set presented in this study could be risky, considering that elevation of AAE is actually dominated by coating (Lack and Langridge, 2013).

4.5 Implications for mixing state

Quantitative direct measurements of BC mixing state and coating thickness are still challenging. SP2 can estimate the coating thickness using a lag-time approach or a Mie calculation approach can be employed, but both methods have a limited range in coating thickness and uncertainties arise from the assumptions made during the retrieval. For example, recent studies found that the mass-equivalent diameter of soot core measured by SP2 could be underestimated due to density assumptions (Y. Zhang et al., 2016). Although size distribution measurement is not available in this study, clues regarding mixing state still can be derived from bulk measurements of optical properties. As discussed in Sect. 4.3, elevated $E_{abs,550}$ observed in the rainy season is associated with aged air masses from a marine origin. To probe the possible mixing state difference between dry and rainy season, $E_{abs,550}$, SSA$_{525}$ and AAE$_{470-660}$ values are used to narrow down the possible core–shell size range as shown in Fig. S17. Monthly averages with 1 SD of AAE$_{470-660}$, SSA$_{525}$ and $E_{abs,550}$ are used as constraints to extract the intersecting core–shell size range from Figs. 2a, S4 and 3a. January and August data are used to represent two different scenarios: elevated AAE$_{470-660}$ (1.19±0.11) with lower $E_{abs,550}$ (1.31±0.32) in dry season and low AAE$_{470-660}$ (1.04±0.09) with elevated $E_{abs,550}$ (1.97±0.71) in rainy season. The results show that January and August have a very different core–shell size range: in January, the core and shell ranges are ∼100–160 nm and ∼120–250 nm, respectively; in August, the core and shell ranges are ∼120–165 nm and ∼170–430 nm, respectively. This confirms again that the soot particles in the rainy season are likely to have a thicker coating than in the dry season.

5 Caveats of the MRS method in its application to ambient data

5.1 Impact of measurement biases

It should be noted that the $E_{abs}$ estimation approach is insensitive to the systematic MAE bias (e.g., systematic over-estimation of $\sigma_{abs}$ and variability of EC mass from different TOA protocols) discussed in Sect. 2.1, because systematic bias in EC mass or $\sigma_{abs}$ is canceled out in the $E_{abs}$ calculation (Eq. 3), since $E_{abs}$ is the ratio of $\sigma_{abs,1}$ to $\sigma_{abs,p}$. To investigate the performance of the MRS approach in response to systematic bias in EC and $\sigma_{abs}$, two simple tests are conducted as shown in Figs. S18 and S19 by adding systematic
biases to the original data. The 1-year measurement data of $\sigma_{abs550}$ and EC are used as original data. Test A represents a situation when $\sigma_{abs}$ is overestimated and EC is underestimated. The biased data are marked as $\sigma'_{abs550}$ and EC', respectively, as shown below:

$$\sigma'_{abs550} = \sigma_{abs550} \times 2,$$  
$$EC' = EC \times 0.7.$$  

(8), (9)

As a result, the average MAE$_{550}$ changed from 18.75 to 53.58 m$^2$g$^{-1}$ and MAE$_p$ changed from 13 to 37 m$^2$g$^{-1}$ (Fig. S18). However, $E_{abs}$ values from the ratio of averages remain the same (1.44).

In Test B, EC from different TOA protocols are compared to investigate the effect of different EC determination approaches while $\sigma_{abs550}$ remains unchanged. EC from IMPROVE TOR protocol is calculated from NIOSH TOT EC following an empirical formula for suburban sites derived from a 3-year OCEC data set in PRD (C. Wu et al., 2016):

$$EC_{IMP\_TOR} = 2.63 \times EC_{NSH\_TOT} + 0.05.$$  

(10)

As shown in Fig. S19, MAE$_{550}$ changed from 18.75 to 7.02 m$^2$g$^{-1}$ and MAE$_p$ changed from 13 to 5 m$^2$g$^{-1}$, but $E_{abs}$ remain almost the same (1.40). The result of Test B implies that although EC is operationally defined, the discrepancy of EC between TOA protocols did not weaken the role of EC in serving as a tracer for primary emissions in MRS application. These examples demonstrate that systematic biases in $\sigma_{abs550}$ and EC have no effects on $E_{abs}$ estimation from the MRS approach.

A study by Cheng et al. (2016) found two distinct types of biomass smoke behave differently on the biases of filter-based $\sigma_{abs}$ measurement. The bias in the first type can be explained by a nearly constant correction factor, which is similar to the situation discussed in Test A. The bias in the second type shows an apparent OC/EC dependence. Test C is carried out to investigate this situation, i.e., examining the impact of sample-dependent bias as a function of $E_{abs}$. Unlike the proportional bias in Test A and B that is the same for all data points, the bias in Test C depends on the $E_{abs550}$ of individual samples, which are parametrized in Eqs. (11) and (12).

$$\sigma'_{abs550} = \sigma_{abs550} + \sigma_{abs550} \times (k \times E_{abs550} - k)$$  
$$EC' = EC - EC \times (k \times E_{abs550} - k)$$  

(11), (12)

As shown in Eqs. (11) and (12), the positive bias of $\sigma_{abs550}$ and negative bias of EC are proportional to $E_{abs550}$. The magnitude of $E_{abs550}$-dependent bias is regulated by the factor $k$.

Since $\sigma'_{abs550}$ and EC' are biased in different directions, resulting in further amplification of MAE biases, which could be considered as the extreme case. As shown in Fig. S20, for $k = 10\%$ (corresponding to a bias of 10% when $E_{abs} = 2$), the bias of MRS-derived $E_{abs}$ is very small (1%). For $k = 20\%$, the MRS-derived $E_{abs}$ changes from 1.44 to 1.66, leading to a bias of 15%. These results imply that if the measurement bias follows the same form as demonstrated in Test C, the bias is not negligible but still acceptable. If the impact only affects $\sigma_{abs}$ or EC rather than impacting both, the bias is expected to be smaller than the estimation shown in Test C.

It should be noted that the parameterization scheme shown in Eqs. (11) and (12) is only for demonstration purposes from a conceptual perspective and it does not necessarily represent real-world measurements. There is a lack of quantitative understanding of this impact. For example, Lee et al. (2007) used artificially fabricated EC samples with OC coatings to evaluate the impact of coating on OC/EC analysis. Biases were observed; nevertheless, the results were linearly correlated with the true OC and EC values with a high $R^2$ (> 0.9), implying that the biases in that specific study were dominated by systematic biases rather than the coating-dependent bias. Further studies are needed to better characterize and parameterize this impact if filter-based techniques are used for $\sigma_{abs}$ and EC determination in the MRS approach.

5.2 Impact of semi-volatile organic carbon

Light absorption contribution due to semi-volatile organic carbon (SVOC) from wood combustion was reported to be negligible in the visible range and around 10–20% at 360 nm (Chen and Bond, 2010). On the other hand, OCEC analysis can be affected by SVOC (Subramanian et al., 2004), either by positive artifacts through adsorption of SVOC onto quartz filters or by negative artifacts through evaporation of SVOC due to the gas–particle re-equilibrium downstream of the VOC denuder. Positive artifacts can be minimized by the installation of a VOC denuder which is widely adopted in RT-OCEC measurements (Ba et al., 2004; Bauer et al., 2009). A typical negative artifact of 10% is expected and can be corrected by backup filters (Subramanian et al., 2004). There was evidence to show that SVOC could affect OC/EC split in thermal–optical analysis (Cheng et al., 2009). However, the bias in EC caused by the OC/EC split drift due to SVOC is systematic, making it fall into the scenario discussed in Test B. As a result, the impact from SVOC on $E_{abs}$ estimation from MRS is expected to be small.

5.3 Impact of mineral dust

The presence of mineral dust (MD) could affect both $\sigma_{abs}$ and EC determination. If MD is externally mixed with soot particles, the light absorption from MD could be miscounted as $\sigma_{abs}$ enhancement, leading to the overestimation of $E_{abs}$. If the light absorption signal from MD is sufficiently strong (e.g., AAE > 2), $\sigma_{abs}$ of MD and BC can be separated with the AAE approach suggested by Fialho et al. (2005). Additionally, the presence of substantial MD in samples has several impacts on the EC determination in thermal analysis. First, if the samples are not pre-treated with acid, the carbon-
ated carbon could be misidentified as EC, resulting in overestimation of EC (Chow et al., 1993). The acid treatment is only available for offline OC/EC analysis and not yet practical for the RT-OC/EC analyzer. Second, metal oxides in MD can lead to premature EC oxidation in the helium stage of OC/EC analysis, leading to underestimation of EC (Wang et al., 2010; Bladt et al., 2012). The lack of a parameterization scheme for correcting the EC loss due to MD makes it improper to use the biased EC as a primary tracer. For these reasons, \( E_{\text{abs}} \) estimation with MRS is not recommended for samples strongly influenced by MD.

5.4 Impact of BrC

The data in this study is dominated by BC absorption that did not show much influence from BrC. However, extra care should be taken if the samples exhibit substantial BrC signature (e.g., AAE > 2). Such situations are equivalent to the two-source scenarios discussed in our previous paper on the MRS method (Wu and Yu, 2016) and the major findings are described below. Two types of two-source scenarios are considered: two correlated primary sources (scenario A) and two independent primary sources (scenario B). In scenario A, in which both BC and primary BrC are dominated by BB, using BC as a solo tracer to calculate the primary ratio \( \text{MAE}_{p, \text{BrC}} \) still works. In scenario B, in which BC and primary BrC are independent, using BC alone to determine a single primary \( \text{MAE}_{p} \) could lead to a considerable bias in \( E_{\text{abs}} \) estimation. Alternatively, if a reliable primary BrC tracer is available, the corresponding \( \text{MAE}_{p, \text{BrC}} \) can be determined with MRS. With the knowledge of \( \text{MAE}_{p, \text{BrC}} \) and \( \text{MAE}_{p, \text{BC}} \), light absorption by BC and BrC can be calculated separately and the \( E_{\text{abs}} \) can be determined using Eq. (13):

\[
E_{\text{abs}} = \frac{\sigma_{\text{abs}, t}}{\sigma_{\text{abs}, p, \text{BC}} + \sigma_{\text{abs}, p, \text{BrC}}} \times \text{MAE}_{p, \text{BC}} \times \text{MAE}_{p, \text{BrC}}.
\]

However, the implementation of Eq. (13) is challenging due to the complexity in the chemical composition of BrC. For example, a recent study found that the 20 most absorbing BrC chromophores account for \( \sim 50 \% \) of BrC light absorption and there is not a single compound contributing more than 10 \% (Lin et al., 2016), making it difficult to choose a single compound as the BrC tracer. In addition, time-resolved measurement of BrC chromophores has yet to be developed. As a result, for scenario B (sample AAE > 2 and primary BrC variations independent of BC), estimation of \( E_{\text{abs}} \) with MRS is not practical at this stage due to the lack of required input data. Using BC alone to determine a single primary \( \text{MAE}_{p} \) could lead to a considerable bias and should be avoided.

6 Conclusions

In this study, a novel statistical approach is proposed and its application on ambient data is demonstrated using 1-year hourly OC and EC data coupled with Aethalometer measurements. Unlike conventional \( E_{\text{abs}} \) determination approaches that require expensive instrumentation (e.g., TD-PAS, VTDMA, SP2), this new approach employs widely deployed instruments (field carbon analyzer and Aethalometer). The key of this new approach involves calculating \( \text{MAE}_{p} \) using the minimum \( R \) squared (MRS) method (Wu and Yu, 2016). The MRS method opens up a new approach to investigate the long-term trend of \( E_{\text{abs}} \) that has rarely been studied using the TD approach. It is found that \( E_{\text{abs}} \) estimation from MRS is insensitive to systematic biases in EC and \( \sigma_{\text{abs}} \) measurements. The annual average \( \text{MAE}_{\text{abs}, 550} \) estimated with MRS is 13 m\(^2\) g\(^{-1}\) and annual average \( \text{MAE}_{\text{abs}, 550} \) is 18.75 ± 6.16 m\(^2\) g\(^{-1}\), suggesting an annual average enhancement factor (\( E_{\text{abs}, 550} \)) of 1.50 ± 0.48 in the PRD region. This value is within the upper limit of \( E_{\text{abs}} \) (~2) in core–shell Mie simulations considering the typical soot size distribution and coating thickness in the PRD.

Both \( \text{MAE}_{\text{abs}, 550} \) and \( E_{\text{abs}} \) show distinct seasonal variations, implying the complexity of soot particle mixing state variations in this region. The elevated summertime \( E_{\text{abs}, 550} \) in the PRD is found to be associated with the domination of aged air masses from the South China Sea, along with the long-range transport of biomass-burning-influenced air masses from Southeast Asia. Core–shell size ranges narrowed down by the \( E_{\text{abs}, 550} \), SSA\(_{525}\) and AAE\(_{470–660}\) constraints suggest that soot particles in the rainy season are likely to have thicker coatings than in the dry season.

Data availability. OC, EC, inorganic ions and \( \sigma_{\text{abs}} \) data used in this study are available from the corresponding authors upon request. The computer programs used for data analysis and visualization in this study are available in Wu (2017a–e).

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

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