Observationally-constrained carbonaceous aerosol source estimates for the Pearl River Delta area of China

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Abstract

We simulated elemental carbon (EC) and organic carbon (OC) aerosols over the Pearl River Delta (PRD) area of China and compared the results to seasonal surface measurements, with the aim of quantifying carbonaceous aerosol sources from a “top-down” perspective. Our regional model was driven by current-best estimates of PRD EC (39.5 Gg C yr\(^{-1}\)) and OC (32.8 Gg C yr\(^{-1}\)) emissions and included updated secondary organic aerosol formation pathways. The simulated annual mean EC and OC concentrations were 4.0 and 7.7 µg C m\(^{-3}\), respectively, lower than the observed annual mean EC and OC concentrations (4.5 and 13.1 µg C m\(^{-3}\), respectively). We used multiple regression to match the simulated EC against seasonal mean observations. The resulting top-down estimate for EC emission in the PRD area was 52.9 ± 8.0 Gg C yr\(^{-1}\). We estimated the OC emission in the PRD area to be 60.2 ± 10.3 Gg C yr\(^{-1}\), based on the top-down EC emission estimate and the primary OC/EC ratios derived from bottom-up statistics. Using these top-down emission estimates, the simulated average annual mean EC and OC concentrations were improved to 4.4 and 9.5 µg C m\(^{-3}\), respectively, closer to the observations. Secondary sources accounted for 42 % of annual mean surface OC in our top-down simulations, with biogenic VOCs being the most important precursors.

1 Introduction

The Pearl River Delta (PRD) area of China (Fig. 1), including parts of the Guangdong Province surrounding the Pearl River estuary (hereafter referred to as the inner PRD), the Hong Kong Special Administrative Region (hereafter referred to as Hong Kong), and the Macau Special Administrative Region (hereafter referred to as Macau), has experienced severe air pollution of fine particulate matter (PM\(_{2.5}\)) in recent years (Chan and Yao, 2008; Liu et al., 2013). In particular, carbonaceous aerosols, including elemental carbon (EC) and organic carbon (OC) aerosols, account for 21–52 % of
the surface PM$_{2.5}$ mass in this area (Cao et al., 2003, 2004; Hagler et al., 2006; Ho et al., 2006; Hu et al., 2012). EC and primary OC (POC) are emitted directly into the atmosphere from incomplete combustion associated with anthropogenic and biomass burning activities. OC is also produced secondarily within the atmosphere (as secondary organic carbon, SOC) from gaseous organic precursors emitted by biogenic, anthropogenic, and biomass burning sources (Hallquist et al., 2009). Accurate quantification of these primary and secondary carbonaceous aerosol sources is paramount to effective air quality management in the PRD area.

Emission inventories are mostly constructed from the “bottom-up” using sectorial activity data and pollutant emission factors. Table 1 summarizes previous bottom-up estimations of PRD carbonaceous aerosol emissions. Total PRD EC emissions from anthropogenic and biomass burning activities were estimated to be 19.4–38.3 and 0.4–0.7 Gg C yr$^{-1}$, respectively. Total PRD OC emissions from anthropogenic and biomass burning activities were estimated to be 28.7–40.9 and 2.1–3.4 Gg C yr$^{-1}$, respectively (Streets et al., 2003b; Zhang et al., 2009; M. He et al., 2011; Zheng et al., 2012). The uncertainties of these bottom-up estimates are considerable, with 95% confidence intervals (CI) ranging from 58 to 495% (Streets et al., 2003b; Zhang et al., 2009; M. He et al., 2011; Zheng et al., 2012).

A few previous studies have used observations to constrain carbonaceous aerosol emissions for China at the national scale (Hakami et al., 2005; Kondo et al., 2011; Fu et al., 2012; X. Wang et al., 2013). Hakami et al. (2005) quantified East Asian black carbon (BC) emissions using BC measurements over South Korea, Japan, and the East China Sea during April 2001. They found that the East Asian domain-wide BC emission agreed well with the estimates of Streets et al. (2003a, b) (143 Gg C during April 2001), but sub-regional emission totals were underestimated for Japan and northeastern China and overestimated for southeastern China. Kondo et al. (2011) analyzed surface BC measurements at an island site in the East China Sea. They estimated Chinese annual BC emission to be 1.92 Tg C yr$^{-1}$, close to the estimate (1.81 Tg C yr$^{-1}$) by Zhang et al. (2009). X. Wang et al. (2013) analyzed BC observations at two Chinese ru-
ral sites, excluding measurements during high pollution events and those impacted by wet deposition. They estimated the Chinese annual BC emission to be 1.80 Tg C yr\(^{-1}\), also similar to the estimate by Zhang et al. (2009). Fu et al. (2012), a forerunner to this current study, used model simulations and surface measurements at 10 Chinese rural and background sites to quantify Chinese annual EC and OC emissions. They estimated EC and OC emissions to be 3.05 ± 0.78 and 6.67 ± 1.30 Tg C yr\(^{-1}\), respectively. To the best of our knowledge, there has not yet been a systematic quantification of PRD carbonaceous aerosol emissions using observational constraints.

Neither the contribution of SOC to total OC, nor the sources and formation pathways of SOC in the PRD area are well understood. Estimates of the secondary fraction of total OC in the PRD area ranged from 9 to 74% with urban/rural and seasonal differences, but there are considerable inconsistencies (Cao et al., 2003, 2004; Ho et al., 2003; Hagler et al., 2006; Gnauk et al., 2008; Hu et al., 2008; L.-Y. He et al., 2011; X.-F. Huang et al., 2011; Hu et al., 2012; Ding et al., 2012). Hu et al. (2008) found that the summertime SOC in Hong Kong were mainly produced by biogenic monoterpenes and sesquiterpenes. In contrast, Ding et al. (2012) found that the SOC at a rural site in central PRD were mostly produced by anthropogenic aromatic precursors year-round. Regional models driven by bottom-up emission inventories for carbonaceous aerosols and organic gaseous precursors found biogenic precursors to be the main source of the simulated SOA (65–90%) in the PRD area (Han et al., 2008; Jiang et al., 2012; Fu et al., 2012). However, these models also tend to underestimate PRD total OC by 30–60% (Han et al., 2008; Jiang et al., 2012; Fu et al., 2012). Most of these models simulated SOA formation by the reversible partitioning of the semi-volatile oxidation products from anthropogenic and biogenic VOC precursors (Pankow, 1994; Odum et al., 1996). More recent modeling studies found that newly discovered pathways, such as the aqueous uptake of dicarbonyls (X.-H. Huang et al., 2011; Fu et al., 2012; Li et al., 2013) and the aging of semi-volatile and intermediate volatility organic compounds (S/IVOCs) (Guo et al., 2014; Matsui et al., 2014) may be important sources of SOC in the PRD area, potentially bridging the gap between simulated and observed OC.
In this study, we used a regional air quality model to simulate seasonal surface EC and OC concentrations in the PRD area for the year 2006. We drove the simulation using current-best bottom-up estimates for PRD carbonaceous aerosol emissions. We compared model results to seasonal surface observations at urban and rural sites to quantify the primary and secondary sources of PRD carbonaceous aerosols from a top-down perspective.

2 Simulations and observations

2.1 The CMAQ model

We used the CMAQ model (version 4.7.1) (Binkowski and Roselle, 2003; Byun and Schere, 2006) to simulate carbonaceous aerosols in the PRD area for the year 2006. CMAQ was driven by simulated meteorological fields from the Weather Research and Forecasting Model (WRF, version 3.2.1) (Skamarock et al., 2008). We simulated three nested domains covering East Asia, Southern China, and the PRD area, with horizontal resolutions of 27, 9, and 3 km, respectively (Fig. S1 in the Supplement). Vertical layers extended from the surface to 50 hPa (26 layers), with 17 layers in the bottom one km. Initial and boundary concentrations of chemical tracers were taken from the monthly mean outputs from a global chemical transport model (Fu et al., 2012). Our simulation was conducted for four 7 day periods in 2006 to represent the typical meteorological and emission conditions for each of the four seasons (4–10 January for winter, 14–20 April for spring, 24–30 July for summer, and 10–16 October for autumn). The model was initialized for the first day of each 7 day period. Results from the second to seventh day were analyzed.

Carbonaceous aerosols in the CMAQ model included EC and POC from anthropogenic and biomass burning emissions (Sect. 2.2), as well as SOC produced from organic gaseous precursors (described below). Carbonaceous aerosols were divided into Aitken and accumulation modes. Microphysical processes were simulated using
moment-based parameterizations (Binkowski and Roselle, 2003). All aerosols were subject to dry and wet deposition. Accumulation and coarse mode aerosols were assumed to be 100% scavenged by cloud and rain water. The wet scavenging efficiency of Aitken mode aerosols in clouds were calculated as a function of air pressure, temperature, and cloud liquid water content (Byun and Schere, 2006).

We included in our CMAQ model three SOC formation pathways. The first two pathways were standard in the CMAQ model: (1) semi-volatile SOC formation by the reversible-partitioning of the semi-volatile oxidation products of VOC precursors (isoprene, monoterpenes, sesquiterpenes, long alkanes, and aromatics) onto preexisting OC (Pankow, 1994; Odum et al., 1996), and (2) non-volatile SOC formation via the oligomerization of semi-volatile SOC (Carlton et al., 2010), via the acid-enhanced oxidation of isoprene, and via low-NOx oxidation of aromatics (Ng et al., 2007; Surratt et al., 2007).

A third SOC formation pathway, involving the irreversible uptake of dicarbonyls by aqueous particles, was implemented into CMAQ and described in detail by Li et al. (2013). Briefly, glyoxal and methylglyoxal were produced by the gas-phase oxidation of isoprene, monoterpenes, propane, >C2 alkenes, acetylene, benzene, toluene, xylenes, phenol, cresols, acetone, glycolaldehyde, and hydroxyacetone. Glyoxal and methylglyoxal were then irreversibly taken up by wet aerosols and cloud droplets with a reactive uptake coefficient $\gamma = 2.9 \times 10^{-3}$ (Liggio et al., 2005; Zhao et al., 2006). The surface area of wet aerosols was diagnosed using composition- and humidity-dependent hygroscopic growth factors (Kim et al., 1993; Whitby and McMurry, 1997). Cloud droplet surface area was diagnosed from the simulated liquid water content in the cloudy fraction of the model grid, assuming effective droplet radii of 10 and 6 µm for maritime and continental clouds, respectively. The Henry’s Law Constants were updated to $4.19 \times 10^5$ M atm$^{-1}$ for glyoxal (Ip et al., 2009) and $3.2 \times 10^4$ M atm$^{-1}$ (Zhou and Mopper, 1990) for methylglyoxal. Li et al. (2013) simulated SOA production during a photochemical smog event in autumn 2009 and compared the results to aerosol mass spectrometry (AMS) measurements at a PRD urban site. They found that including the
dicarbonyl SOA pathway significantly improved the agreement between the simulated SOA and the observed oxygenated organic aerosol (OOA), both in terms of the concentrations and the correlations with sulfate (a secondary aerosol mainly produced in the aqueous phase).

Our model did not include SOC formation from the aging of S/IVOCs (Robinson et al., 2007). Recent modeling studies showed that including this aging process may increase the simulated SOA concentrations by approximately 0.5 \( \mu \text{g m}^{-3} \) over the PRD area in autumn (Guo et al., 2014) and by approximately 3 \( \mu \text{g m}^{-3} \) over southern China in spring (Matsui et al., 2014). Thus, our simulated SOC (Sect. 5) is likely biased low and we will discuss the impacts in Sect. 5. However, this potential underestimation of SOC does not affect our top-down quantification of primary carbonaceous aerosol emissions.

### 2.2 Emissions of EC, OC, and VOC precursors

We combined four current-best bottom-up emission inventories of EC, OC, and VOC precursors for the PRD area and the rest of East and South Asia to drive our simulation. Tables 2 and 3 summarize the bottom-up annual emissions estimates of carbonaceous aerosols and VOC precursors for the PRD area used in our simulation.

Anthropogenic EC and OC emissions for the inner PRD were taken from the inventory developed by Zheng et al. (2012) for the year 2009, which had a native resolution of 3 km and included emissions from power generation, industry, transportation, and residential activities. Activity data were collected from provincial yearbooks. Emission factors for residential combustion were obtained from Chinese measurements. EC and OC emissions from other sectors were estimated using PM\(_{2.5}\) emission factors and EC and OC fractional abundance in PM\(_{2.5}\). Anthropogenic EC and OC emissions for Hong Kong, Macau, and the rest of East and South Asia were taken from the inventory developed by Zhang et al. (2009) for the year 2006, which had a native resolution of 0.5°. We interpolated their emission fluxes to our nested domains. Emission factors were determined based on a combination of Chinese measurements and data from western studies. The residential emissions from the Zheng
et al. (2012) inventory were misrepresented in space. We spatially distributed the residential emissions from Zheng et al. (2012) using the spatial pattern from the inventory developed by Zhang et al. (2009). We applied diurnal (Q. Zhang, personal communication, 2012) and seasonal variations (Zheng et al., 2012; Zhang et al., 2009) to the PRD and Chinese anthropogenic emissions based on activity statistics. No seasonal and diurnal variations were applied to anthropogenic emissions outside China. The total annual anthropogenic EC emission for the PRD area was 38.9 Gg C yr$^{-1}$, including 25.7 Gg C yr$^{-1}$ from transportation, 7.2 Gg C yr$^{-1}$ from industry, 4.9 Gg C yr$^{-1}$ from residential sources, and 1.1 Gg C yr$^{-1}$ from power generation. Annual anthropogenic OC emission for the PRD area was 29.8 Gg C yr$^{-1}$, including 12.9 Gg C yr$^{-1}$ from transportation, 12.3 Gg C yr$^{-1}$ from residential sources, 3.8 Gg C yr$^{-1}$ from industry, and 0.8 Gg C yr$^{-1}$ from power generation. For both EC and OC, Hong Kong and Macao combined only contributed approximately 2 % of the total emissions from the entire PRD area. The uncertainties (95 % CI) for residential emissions in the inner PRD were 2.0–10.9 Gg C yr$^{-1}$ for EC and 5.5–24.8 Gg C yr$^{-1}$ for OC (Zheng et al., 2012). The uncertainties for other sectorial emissions in the inner PRD, while likely to be high, were not quantified in Zheng et al. (2012). Lei et al. (2011) analyzed the uncertainties of their EC and OC estimates for China. We used here their uncertainties (95 % CI) for industrial (±300 % for EC and ±365 % for OC), power generation (±50 % for EC and ±36 % for OC), and transportation (±91 % for EC and ±70 % for OC) sources for the inner PRD.

Biomass burning EC and OC emissions within the inner PRD were taken from the inventory developed by M. He et al. (2011) for the year 2006. The inventory had a resolution of 3 km and included emissions from field burning of agricultural residues and from wild fires. Activity data were from Chinese statistics. Emission factors were specific to Chinese and Asian biomass types (Cao et al., 2008; Li et al., 2009; Chang and Song, 2010). Biomass burning EC and OC emissions in Hong Kong, Macau, and the rest of East and South Asia were taken from the inventory by Streets et al. (2003b), representative of mid-1990s burning conditions at a native resolution of 1°. We interpo-
lated the emission fluxes to our nested domains. We applied seasonal variations to the inner PRD biomass burning emissions based on activity statistics (M. He et al., 2011). No seasonal variations were applied to biomass burning emissions outside of the inner PRD. A few Asian biomass burning emission inventories have been developed for more recent years (Song et al., 2010; Huang et al., 2012). However, since the PRD EC and OC measurements incorporated in this study were not sensitive to biomass burning sources outside of the PRD area (as will be shown in Sect. 4.1), the choice of biomass burning emission inventory outside of the PRD area does not heavily influence our top-down emission estimates (see discussion in Sect. 4.1). Annual biomass burning EC emission in the PRD area was 0.6 Gg C yr\(^{-1}\), including 0.4 Gg C yr\(^{-1}\) from field burning of rice straw and 0.2 Gg C yr\(^{-1}\) from field burning of other residues. Annual biomass burning OC emission in the PRD area was 3.0 Gg C yr\(^{-1}\), including 1.4 Gg C yr\(^{-1}\) from field burning of rice straw, 1.4 Gg C yr\(^{-1}\) from field burning of other residues, and 0.2 Gg C yr\(^{-1}\) from forest fires. The 95 % CI for biomass burning source in inner PRD were 0.1–1.2 Gg C yr\(^{-1}\) for EC and 1.1–8.4 Gg C yr\(^{-1}\) for OC (M. He et al., 2011). Combined with anthropogenic emissions, we estimated the overall emission uncertainties (95 % CI) in the inner PRD to be ±161 % for EC and ±122 % for OC (see Table 2).

Figures 2 and 3 show the spatial distributions of the annual EC and OC emissions for the PRD area from the bottom-up inventories used to drive our simulation. For each sector, the spatial patterns of EC and OC emissions were similar owing to their common combustive origin. Industrial and power generation emissions were dotted in the inner PRD, reflecting the locations of factories and power plants. Transportation emissions were highest over downtown Guangzhou and Shenzhen, which are the two most populated cities in the PRD area. Biomass burning emissions were mostly over the rural agricultural areas around Guangzhou, Dongguan, Jiangmen, Huizhou, and Zhaoqing, reflecting the burning of crop residue for land clearing purposes (mostly in winter) (M. He et al., 2011).

Table 3 summarizes the PRD VOC emissions we used to drive our model. For the inner PRD, anthropogenic and biomass burning VOC emissions were taken from the
inventories developed by Zheng et al. (2009) and M. He et al. (2011), respectively. For Hong Kong, Macau, and the rest of East and South Asia, anthropogenic and biomass burning VOC emissions were taken from the inventories developed by Zhang et al. (2009) and Streets et al. (2003b), respectively. Biomass burning emissions of glyoxal, methylglyoxal, glycolaldehyde, and hydroxyacetone were scaled from biomass burning CO emissions following Fu et al. (2008). Biogenic VOC emissions were calculated using the MEGAN algorithm (Guenther et al., 2006), driven by meteorological fields simulated by WRF and land cover and vegetation data from MODIS. Annual isoprene and monoterpenes emissions over the PRD area were 221 and 157 Gg yr$^{-1}$ respectively. Figure 4 shows the spatial distributions of anthropogenic xylenes, biomass burning glyoxal, and biogenic isoprene emissions. Anthropogenic xylenes emissions were highest near city centers and along major highways, reflecting vehicular sources. Biomass burning glyoxal emissions were mostly over rural agricultural areas, reflecting the burning of crop residue. Biogenic isoprene emissions were highest over the forested areas in the northwestern and northeastern parts of the PRD area.

2.3 Seasonal surface measurements of carbonaceous aerosols

We needed seasonal carbonaceous aerosol measurements with sufficient spatiotemporal coverage and resolution to provide top-down constraints on the carbonaceous aerosol sources in the PRD area. There have been many individual, short-term carbonaceous aerosol measurements in the PRD area, but data from a single, consistent network of long-term measurements is not yet available. We compiled a database consisting of seasonal carbonaceous aerosol measurements from 11 studies conducted at 23 PRD locations, including 15 urban (non-roadside) locations and 8 rural locations, between the years 2000 and 2008 (Cao et al., 2003, 2004; Ho et al., 2003; Hagler et al., 2006; Gnauk et al., 2008; Hu et al., 2008; L.-Y. He et al., 2011; Hu et al., 2012; X.-F. Huang et al., 2011; K. F. Ho, unpublished data, 2012; J. Tao, unpublished data, 2012). Figure 1 shows the spatial distribution of the 23 locations. Table 4 lists the details of each location. The measurements compiled here involved a number of different
sampling and analysis techniques, but there is currently no way to compare them in a consistent manner. We summarized the techniques used in Table 4 (additional information in Sect. S1 in the Supplement) and noted that there may be inconsistencies.

To reflect the seasonal variation, we averaged the measurements for each of the four seasons: spring (March–May), summer (June–August), autumn (September–November), and winter (December, January, and February). Carbonaceous aerosol measurements were available at 9 locations in Hong Kong, including 6 urban locations (BU, CW, KT, YL, UST, and TW) and 3 rural locations (TC, HT, and TM). However, our model was unable to resolve the spatial variability due to the low resolution of the bottom-up emission inventory in Hong Kong. We averaged the seasonal measurements at the 9 locations in Hong Kong to represent hereafter a single urban site, HK. Table 5 summarizes the number of available measurements of urban and rural sites in the PRD area in each season. A total of 43 seasonal measurements (32 urban and 11 rural) were used in Sect. 4 to quantify PRD carbonaceous aerosol sources.

Table 4 shows the seasonal surface EC and OC concentration measurements incorporated in our database. The seasonal mean surface EC and OC concentrations in all urban and rural sites varied in the range of 1.4–13.3 and 5.7–38.4 µgCm$^{-3}$, respectively. These concentrations are similar to the seasonal means typically observed in Shanghai (1–33 µgCm$^{-3}$ for EC and 2–27 µgCm$^{-3}$ for OC; Ye et al., 2003; Feng et al., 2006, 2009; Chen et al., 2008; Hou et al., 2011; Cao et al., 2012, 2013; Geng et al., 2013) and Beijing (2–32 µgCm$^{-3}$ for EC and 6–41 µgCm$^{-3}$ for OC; He et al., 2001; Dan et al., 2004; Sun et al., 2004; Duan et al., 2006; Zhou et al., 2012; Zhang et al., 2013; Zhao et al., 2013). The highest seasonal mean surface concentrations were observed in winter for both EC and OC at all sites.

For the purpose of constraining OC sources, it is useful to quantify the POC and SOC concentrations in the observed total OC. However, there is currently no way to unambiguously separate the primary or secondary components in OC or OA measurements. Seven of the 11 measurement studies incorporated in our database quantified
POC and SOC using the EC tracer method:

$$\text{SOC} = \text{OC} - \text{POC} = \text{OC} - \text{EC} \times (\text{POC}/\text{EC})\text{.}$$  \hspace{1cm} (1)$$

The POC/EC ratio in Eq. (1) can be estimated using several methods. Most studies adopted the seasonal or annual minimum observed OC/EC ratio as the POC/EC ratio (Cao et al., 2003, 2004; K. F. Ho unpublished data, 2012; J. Tao unpublished data, 2012), assuming that there was no SOC when the observed OC/EC ratio was at this minimum value. Alternatively, Hu et al. (2012) estimated the POC/EC ratio using the OC/EC value corresponding to a minimum correlation between SOC and EC. Both methods assumed that the POC/EC ratio remained constant during the period of observation, which may not be true if emission sources change.

Two AMF studies were incorporated in our database (X.-F. Huang et al., 2011; L.-Y. He et al., 2011). These studies applied the positive matrix factorization (PMF) technique on the high-resolution mass spectra and identified four OA components, including hydrocarbon-like OA (HOA), biomass burning OA (BBOA), semi-volatile oxygenated OA (SV-OOA), and low-volatile oxygenated OA (LV-OOA). We used the total oxygenated OA (SV-OOA + LV-OOA) as a surrogate for total SOA. We estimated the SOC mass using the OC/OM mass ratios observed by L.-Y. He et al. (2011) and X.-F. Huang et al. (2011). Finally, Hu et al. (2008) estimated total SOC mass using known SOA tracers from isoprene, monoterpenes, $\beta$-caryophyllene, and toluene.

Table 4 summarizes the seasonal mean surface SOC concentrations measured at 15 urban and rural sites. SOC concentrations ranged from 1.2–19.1 µg C m$^{-3}$ for urban sites and from 2.0–6.7 µg C m$^{-3}$ for rural sites. At all sites where observations were available, the fraction of SOC in total OC were higher in summer (57%) and lower in winter (46%), likely due to the stronger photochemistry and the higher biogenic precursor emissions in warmer seasons.
3 Simulated carbonaceous aerosols with bottom-up emissions

Figure 5 shows our simulated surface EC concentrations using the bottom-up emissions and compares them against the observations. The simulated annual mean EC concentration averaged over the urban and rural sites (sampled at sites where seasonal observations were available) was 4.0 µg C m\(^{-3}\), 11 % lower than the observed annual mean concentration (4.5 µg C m\(^{-3}\)). Simulated seasonal EC concentrations ranged from 0.9–8.3 µg C m\(^{-3}\) and were highest in winter and lowest in summer, consistent with the observations. The gap between simulated and observed seasonal EC concentrations were within ±2.0 µg C m\(^{-3}\) for rural sites and within ±3.5 µg C m\(^{-3}\) for urban sites, except at three industrial sites: HP, LG, and LH. The prevailing low-level wind over the PRD in autumn and winter is northeasterly, which transports regional pollutants from inner Mainland China to the PRD. Our model reproduced well the observed EC concentrations at the CH site at northern edge of the PRD area in autumn and winter, with biases less than ±0.7 µg C m\(^{-3}\). In spring and summer, the prevailing low-level wind over the PRD is southeasterly (from the Pacific). The model was also able to reproduce the relatively lower EC concentrations at the HK site in spring and summer. These show that the model reasonably captured the regional EC background transported from inner Mainland China or from the Pacific into the PRD domain.

Figure 5 also shows the sectorial contributions to the simulated surface EC based on sensitivity simulations using the bottom-up emissions. Simulations using the bottom-up emission estimates indicated transportation to be the largest annual mean EC source for most urban (averaged 58 %) and rural (averaged 53 %) sites. The second largest annual mean EC source was industrial emissions (13 %), followed by residential emissions (9 %). Contributions to the annual mean surface EC from power generation (3 %) and biomass burning (1 %) were minor. The contribution of annual mean EC transported from outside of the PRD area averaged 17 % for all urban and rural sites.

Figure 6 shows our simulated surface OC concentrations using the bottom-up emissions and compared them against observations. The simulated annual mean OC con-
Concentration averaged over urban and rural sites (sampled at sites where seasonal observations were available) was 7.7 µg C m⁻³, 41% lower than the observed 13.1 µg C m⁻³. Simulated seasonal OC concentrations ranged from 3.0–16.0 µg C m⁻³ and were highest in autumn and lowest in summer, consistent with the observations. The simulated seasonal OC concentrations agreed well with observations at rural sites, with biases less than ±6.5 µg C m⁻³ and a normalized mean bias (NMB) of −20%. However, the simulated seasonal OC concentrations were significantly lower than observations at most urban sites (biases up to ±28 µg C m⁻³, NMB = −45%), except at SU and ZQ (NMB = −16%). Our model reproduced well the observed POC concentrations at the CH site in autumn and winter (biases less than ±1.1 µg C m⁻³) and at the HK site in spring and summer (biases less than ±1.1 µg C m⁻³). These indicate that our model was able to capture the regional POC background transported from outside of the PRD area.

Figure 6 also shows the primary sectorial and secondary contributions to surface OC concentrations. Using the bottom-up emission, transport from outside of the PRD area was the most important primary OC source, contributing 17% of the simulated annual mean surface OC at urban sites and 24% of that at rural sites. PRD transportation was the second largest primary OC source, contributing 17% averaged for all urban and rural sites, while other PRD sources (biomass burning, industry, power generation, and residential sources) combined to contribute 18%. Moreover, SOC was an important component of PRD OC in all seasons, contributing 43 and 52% of the simulated annual urban and rural mean surface OC, respectively. The contribution of SOC was particularly pronounced in summer (55%) and autumn (50%). Even in winter, SOC contributed 32% of the simulated seasonal urban and rural mean surface OC.

Figure 7 shows the scatter plots of simulated vs. observed seasonal mean surface EC, OC, and POC concentrations at PRD sites. Using the bottom-up emission estimates, our simulations underestimated the EC seasonal mean observations (reduced major axis regression slope = 0.79), with moderate performance at capturing the spatiotemporal variability (correlation coefficient r = 0.54). For OC and POC, our model
also underestimated the seasonal mean observations, with regression slopes of 0.41 and 0.57, respectively. The model had little skill in capturing the spatiotemporal variability of the seasonal mean OC and POC observations ($r = 0.35$ for OC and $r = 0.44$ for POC).

4 Top-down estimates of carbonaceous aerosol emissions

4.1 Top-down EC emission estimates

Using observations as constraints, we derived top-down estimates on EC sources in the PRD area. We divided the sources of PRD surface EC into four parts: (1) anthropogenic transportation, (2) anthropogenic non-transportation (including industrial, power generation, and residential sources), (3) biomass burning, and (4) background EC transported from outside of the PRD area. The reason for representing the industrial, power generation, and residential sources as one combined non-transportation source was that the simulated EC contributions from these sources showed high spatial correlations ($r = 0.49–0.63$) and the observations used in this study were not able to distinguish them. As discussed in Sect. 3, our model satisfactorily reproduced the seasonal regional background EC levels (EC transported from outside of the PRD area). Hence, we assumed that the simulated regional EC background levels were reasonable, and subtracted them from the seasonal observations. We used multiple regression to match the simulated EC contributions from each sector against seasonal mean observations at the 15 urban and rural sites. The multiple regression model is:

$$c_{\text{obs}} - c_{\text{background}} = \beta_1 c_{\text{transportation}} + \beta_2 c_{\text{non-transportation}} + \beta_3 c_{\text{biomass}} + \varepsilon. \quad (2)$$

In Eq. (2), $c_{\text{obs}}$ is the vector of observed seasonal EC concentrations. $c_{\text{background}}$ is the vector of simulated regional EC background contributed by sources from outside of the PRD area. $c_{\text{transportation}}$, $c_{\text{non-transportation}}$, and $c_{\text{biomass}}$ are the vectors of simulated EC contributed from anthropogenic transportation, from anthropogenic non-transportation
(industrial, power generation, and residential sources), and from biomass burning sources, respectively. $\beta_1$, $\beta_2$, and $\beta_3$ are the domain-wide scalar scale factors required for these sources to best match the seasonal observations obtained from the multiple regression. $\mathbf{e}$ is the regression model error vector. We further assumed that the bottom-up estimate for EC emissions from biomass burning activities was correct, setting $\beta_3 = 1$ and subtracting the simulated $\beta_3 \mathbf{c}_{\text{biomass}}$ from the right-hand-side of Eq. (2). The rationale for this assumption is two-fold. Firstly, the bottom-up estimate for biomass burning EC emissions was two orders of magnitude smaller than the emissions from anthropogenic activities (Table 2). Secondly, at least at the time and locations where EC measurements were available, our simulation using the bottom-up inventories showed that biomass burning contributed very little to surface EC concentrations (annual-average 1 % for all sites). The scale factor for transportation emissions ($\beta_1$) obtained from the multiple regression was 0.49, with a standard deviation of $\pm 0.12$ and a $p$ value of 0.007. The scale factor for the anthropogenic non-transportation sector ($\beta_2$) was 3.01, with a standard deviation of $\pm 0.56$ and a $p$ value < 0.001. The multiple regression explained 78 % of the observed variance (adjusted $R^2 = 0.78$).

Table 2 summarizes our top-down EC emission estimates. The estimated total EC emission from the PRD area was $52.9 \pm 8.0 \text{ Gg C yr}^{-1}$, including $39.7 \pm 7.4 \text{ Gg C yr}^{-1}$ from anthropogenic non-transportation sources, $12.6 \pm 3.1 \text{ Gg C yr}^{-1}$ from anthropogenic transportation source, and $0.6 \pm 0.8 \text{ Gg C yr}^{-1}$ from biomass burning source (unchanged from the bottom-up biomass burning emission inventory). The corresponding 95 % CI of total top-down estimated EC was $\pm 30\%$, calculated assuming normality for emission estimate probabilities. Our top-down total EC emission estimate is 34 % larger than the bottom-up estimate (39.5 Gg C yr$^{-1}$).

4.2 Top-down OC emission estimates

We used two methods to estimate the PRD OC emissions. Firstly, we used the multiple regression technique to best-match the simulated POC against seasonal POC
observations (Table 4). The multiple regression model is:

\[ c_{\text{obs}} - c_{\text{background}} = \gamma_1 c_{\text{transportation}} + \gamma_2 c_{\text{non-transportation}} + \gamma_3 c_{\text{biomass}} + \epsilon. \]  

(3)

In Eq. (3), \( c_{\text{obs}} \) is the vector of the observed seasonal POC concentrations. \( c_{\text{background}} \) is the vector of simulated regional POC background contributed by sources from outside the PRD area. \( c_{\text{transportation}} \), \( c_{\text{non-transportation}} \), and \( c_{\text{biomass}} \) are the vectors of the simulated POC contributions from anthropogenic transportation, from anthropogenic non-transportation, and from biomass burning sources, respectively. \( \gamma_1 \), \( \gamma_2 \), and \( \gamma_3 \) are the corresponding scalar scale factors for these sources. \( \epsilon \) is the regression model error vector. We assumed that the simulated POC from the outside PRD and biomass burning sources were correct, for reasons similar to the case of EC (as discussed in Sect. 3). The multiple regression gave scale factors \( \gamma_1 = 0.55 \) and \( \gamma_2 = 2.88 \). The resulting top-down estimate for PRD OC emission was 58.8 Gg C yr\(^{-1}\).

Alternatively, we used a “hybrid” approach to estimate the OC emissions. We used the coefficients \( \beta_1 \), \( \beta_2 \) and \( \beta_3 \) from the multiple regression analysis for EC emission in Eq. (2) to scale the bottom-up OC emission estimates. This is equivalent to combining the POC/EC emission ratios from bottom-up inventories with the observed EC constraints. The resulting estimated total OC emission was 60.2 Gg C yr\(^{-1}\).

The estimated PRD OC emissions using the two methods were similar. We chose the second (hybrid approach) estimate for further discussion, because its uncertainties come from two traceable origins: (1) explicitly from the multiple regression analyses on EC observations, and (2) implicitly from the POC/EC emission ratios used in the bottom-up inventories. In contrast, the uncertainty of the first estimate depends on the uncertainty of estimating POC concentrations from total OC measurements, which is much harder to quantify. Using the hybrid approach, the total estimated OC emission in the PRD area was 60.2 ± 10.3 Gg C yr\(^{-1}\), including 50.9 ± 9.5 Gg C yr\(^{-1}\) from anthropogenic non-transportation sources, 6.3 ± 1.5 Gg C yr\(^{-1}\) from anthropogenic transportation source and 3.0 ± 3.8 Gg C yr\(^{-1}\) from biomass burning (unchanged from the bottom-up biomass burning emission inventory). The uncertainty (estimated based on
the 95% CI of the EC multiple regression) of total top-down estimated OC was ±34%. The top-down estimated total OC emission estimate was 84% larger than the bottom-up estimate (32.8 Gg C yr\(^{-1}\)).

Figure 8 shows the spatial distributions of the top-down EC and OC emission estimates. EC and OC emissions were enhanced over the entire PRD domain, with largest emissions over the center of the PRD area. The difference between the top-down and the bottom-up emissions was mainly due to the increase of the anthropogenic non-transportation sources.

### 4.3 Evaluation of the top-down emission estimates

We simulated seasonal surface EC and OC using our top-down estimates of carbonaceous aerosol emissions to evaluate their performance. Figure 5 shows the spatial and temporal distributions of the simulated seasonal mean surface EC concentrations using the top-down emissions and compared them to the observations. Our model performance improved for most sites at all seasons (averaged NMB from −13 to −2%) compared to the simulation using the bottom-up emissions. Simulated EC increased significantly at industrial sites (HP and LG), correcting the low-biases there. Simulated seasonal EC concentrations at downtown sites (at LW in spring, at SU in summer, at SCI in autumn, and at PKU in all seasons) decreased and were also closer to the observations. The simulated annual mean EC concentration using top-down emissions for urban and rural sites (sampled at sites where seasonal observations were available) was 4.4 µg C m\(^{-3}\), only 3% lower than the observed 4.5 µg C m\(^{-3}\).

Figure 5 also shows the sectorial contributions to the simulated EC using our top-down emission estimates. Anthropogenic non-transportation activities became the largest source of EC at most sites year-round, contributing on average 62% for urban sites and 56% for rural sites. For the industrial sites HP and LG, the contributions of anthropogenic non-transportation sources increased significantly (79% for HP and 57% for LG) compared to the simulation using the bottom-up inventories (41% for HP and 20% for LG).
Figure 6 shows the simulated seasonal mean surface OC concentrations using the top-down emissions and compared them against the observations. The simulated OC concentrations increased at all sites year-round and were closer to the observations. The simulated annual mean OC for urban and rural sites (sampled at sites where seasonal observations were available) was 9.5 µgC m⁻³, 27% lower than the observed 13.1 µgC m⁻³. Anthropogenic non-transportation sources were the main sources of OC for most sites year-round, contributing on average 42% for urban sites and 30% for rural sites. The contributions of anthropogenic non-transportation sources at the industrial sites (HP and LG) also increased significantly (60% for HP and 37% for LG) compared to the simulation using the bottom-up inventories (31% for HP and 15% for LG).

Figure 9 shows the spatial distributions of the simulated annual mean surface OC concentrations using the top-down emissions. The total simulated annual mean surface OC over the entire PRD domain was 7.9 µgC m⁻³. Anthropogenic emissions of OC was the largest source of surface OC in our simulation, contributing 38% (3.0 µgC m⁻³) of the annual mean surface OC. Primary OC transported from outside of the PRD area contributed 18% of the annual mean surface OC (1.4 µgC m⁻³). Biomass burning source played a minor role in the annual mean surface OC (3%, 0.2 µgC m⁻³).

Figure 7 compares the overall model performances using the bottom-up and top-down simulations. Using the top-down EC emission estimates, the model’s ability at reproducing the magnitude and the spatiotemporal variability of the EC observations were significantly improved (slope = 0.90, r = 0.65). The model’s performance at reproducing the magnitudes and spatiotemporal variability of POC and total OC were also improved using the top-down OC emission estimates, although to smaller extents (slope = 0.79 for POC and 0.51 for total OC; r = 0.56 for POC and 0.50 for total OC).

We compared our top-down estimated carbonaceous aerosol emissions in the PRD area with previous studies (Table 1). Our anthropogenic EC emission estimate is 170% larger than the estimation of Zhang et al. (2009) (19.4 GgC yr⁻¹) and 37% larger than that of Zheng et al. (2012) (38.3 GgC yr⁻¹). Our anthropogenic OC (53.6 GgC yr⁻¹) is
40 % larger than the estimation of Zhang et al. (2009) (40.9 Gg C yr\(^{-1}\)) and 99 % larger than that of Zheng et al. (2012) (28.7 Gg C yr\(^{-1}\)).

The gap of these individual bottom-up emission inventories could be caused by two possible reasons. First, previous studies usually like to adopt the middle value of emission factors from results of western measurements when the local data were not available. Zhang et al. (2009) and Zheng et al. (2012) estimated EC and OC emission factors in power generation, industrial, and mobile sources by PM\(_{2.5}\) emission factors and mass ratios of EC and OC in PM\(_{2.5}\). The PM\(_{2.5}\) emission factors were based on a combination of domestic and western measurements (Zhang et al., 2007; Zhao et al., 2010), but the mass ratios were mostly from foreign studies (Bond et al., 2004; Kupiainen et al., 2004). However, previous aerosol source profile studies showed out that the mass ratios of EC and OC in Chinese PM\(_{2.5}\) are often larger from those in western studies (He et al., 2006, 2008; Wang et al., 2009). Another reason may be that previous bottom-up estimates are generally considered to underestimate activity levels and/or overestimate emission control technology. For example, the removal efficiencies of BC and OC in power generations and industrial combustions were assumed the same as that of PM\(_{2.5}\) in Zhang et al. (2009) and Zheng et al. (2012), although the efficiencies of submicron particles are expected to be lower than larger particles.

5 Secondary sources of OC in the PRD area

Finally, we quantified the secondary contributions to surface OC in the PRD area. Our total simulated annual mean surface SOC concentration averaged over 15 urban and rural sites was 3.5 µg C m\(^{-3}\), 51 % lower than the observed 7.2 µg C m\(^{-3}\). The model performance at simulating SOC was best in autumn (NMB = −12 %), followed by spring (NMB = −53 %), summer (NMB = −62 %), and winter (NMB = −78 %). The model performance at simulating annual mean SOC was better for rural sites (NMB = −32 %) than for urban sites (NMB = −57 %). Figure 6 shows that the simulated total OC using the top-down emission estimates agreed well with the observations at rural site.
but still underestimated observations at urban sites. Figure 7 shows that this gap was mainly owing to low-biases in our simulated urban SOC. These findings indicate that our model may still be underestimating SOC from anthropogenic sources. Our model successfully reproduced the observed SOC for rural sites especially in warmer seasons (NMB = 0% for autumn and −40% for summer). This indicates that our model simulated the SOC from biogenic source reasonably well.

Figure 9 shows the contributions of different SOC formation pathways in our model. Semi- and non-volatile SOC formation combined contributed 1.7 µgC m$^{-3}$ to the annual mean surface OC. Aqueous uptake of dicarbonyl similarly contributed 1.6 µgC m$^{-3}$ to the annual surface mean OC. Both types of SOC were higher in the north-west of the PRD area, reflecting the impact of high biogenic VOC emissions. In addition, semi- and non-volatile SOC concentrations also had high values over city centers, reflecting the impact of anthropogenic VOC precursor emissions. Combined, secondary sources accounted for 42% of annual mean surface OC over the PRD area in our model. The result is similar with previous modeling studies (30–40%) (Jiang et al., 2012; Han et al., 2008) but potentially lower than observation constraints (35–74%) (Cao et al., 2003, 2004; Ho et al., 2003; Hagler et al., 2006; Gnaauk et al., 2008; Hu et al., 2008; L.-Y. He et al., 2011; Hu et al., 2012; X.-F. Huang et al., 2011; K. F. Ho, unpublished data, 2012; J. Tao, unpublished data, 2012) in the PRD area.

Table 3 summarizes the sources of surface SOC as simulated by our model using the top-down emission estimates. Biogenic VOC was the dominant source of SOC in our model, contributing 74% of the simulated annual mean SOC. Isoprene and monoterpenes each contributed approximately half of the SOC from biogenic sources. Anthropogenic and biomass burning VOCs accounted for 25 and 1% of the simulated annual mean SOC, respectively. Hu et al. (2008) found monoterpenes and β-caryophyllene to be the largest sources of summertime SOA in Hong Kong, while Ding et al. (2012) found aromatics accounted for 76–79% of the identified SOC in central PRD in summer, autumn and winter. Both Hu et al. (2008) and Ding et al. (2012) were based on SOA tracer analyses. Using an emission-based box model, S. Wang et al. (2013)
showed that the sources of SOC in the PRD area was highly variable across locations and seasons due to the spatiotemporal variability of precursor emissions. They found monoterpenes to be the most important contributor in Hong Kong, while toluene and xylenes were more important in Guangzhou.

As discussed above, our model likely underestimated SOC from anthropogenic sources. This may be in part due to the omission of SOC formation via the aging of S/IVOCs (Guo et al., 2014; Matsui et al., 2014) in our model. Matsui et al. (2014) simulated OA in East Asia including the SOA formation from S/IVOCs. They found that including the aging processes of gaseous organic precursors (alkanes, olefins, aromatics, isoprene, monoterpenes, and sesquiterpenes) could narrow, but not close, the gap between simulated and observed SOA in southern China (simulated SOA increased from 1 to 4 µg m⁻³, but was still underestimated by 30–40 %) and greatly enhance the interaction of anthropogenic and biogenic sources (biogenic SOA enhanced by 45 %). However, Matsui et al. (2014) still found biogenic sources to be the largest contributor of SOA over southern China, consistent to our results.

6 Conclusions

We used a regional chemical model to simulate seasonal carbonaceous aerosol concentrations in the PRD area of China and compared the results to surface measurements, with the aim of quantifying PRD carbonaceous aerosol sources. We used a combination of several current bottom-up emission estimates for EC (39.5 Gg C yr⁻¹) and OC (32.8 Gg C yr⁻¹) to drive our simulation. SOC formation in our model takes place via reversible-partitioning of semi-volatile oxidation products from VOCs, non-volatile SOC formation, and aqueous uptake of dicarbonyls.

Using the bottom-up emission inventories, the simulated results were 11 and 41 % lower than the surface EC and OC observations at 15 urban and rural sites, respectively. We used multiple regression to match the simulated EC contributions from different sectors against seasonal mean observations at all urban and rural sites. The re-
resulting top-down estimate for PRD EC emissions is $52.9 \pm 8.0 \text{ Gg C yr}^{-1}$. We combined the POC/EC ratios from bottom-up statistics and our top-down EC source estimate to constrain OC emissions. The resulting top-down estimate for PRD OC emission is $60.2 \pm 10.3 \text{ Gg C yr}^{-1}$. The agreement between model results and observations is significantly improved in the simulation using our top-down emissions, compared to that using the bottom-up emissions. The simulated annual mean EC concentration using top-down emission averaged over all urban and rural sites is $4.4 \mu g \text{ C m}^{-3}$, 3% lower than the observed $4.5 \mu g \text{ C m}^{-3}$. The simulated annual mean OC concentration using top-down emission averaged over all urban and rural sites is $9.5 \mu g \text{ C m}^{-3}$, 27% lower than the observed $13.1 \mu g \text{ C m}^{-3}$. Applying the top-down emissions, the model is also able to better capture the observed seasonal EC and OC concentrations and their spatiotemporal variability.

Secondary sources account for 42% of annual mean surface OC in our top-down simulation. The most important source for SOC in the PRD area is biogenic emissions (74%), with isoprene (37%) and monoterpenes (36%) being the major precursors. Our findings are important for the understanding and management of air pollution in the PRD area.

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References


Zhang, R., Jing, J., Tao, J., Hsu, S.-C., Wang, G., Cao, J., Lee, C. S. L., Zhu, L., Chen, Z., Zhao, Y., and Shen, Z.: Chemical characterization and source apportionment of PM$_{2.5}$ in
Table 1. Comparison of recent estimates for carbonaceous aerosol emissions in the PRD area.

<table>
<thead>
<tr>
<th>Year</th>
<th>Year</th>
<th>EC (Gg C yr(^{-1}))</th>
<th>OC (Gg C yr(^{-1}))</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Anthropogenic</td>
<td>Biomass burning</td>
<td>Anthropogenic</td>
</tr>
<tr>
<td>Streets et al. (2003b)</td>
<td>2000</td>
<td>–</td>
<td>0.7</td>
</tr>
<tr>
<td>Zhang et al. (2009)</td>
<td>2006</td>
<td>19.4</td>
<td>–</td>
</tr>
<tr>
<td>M. He et al. (2011)</td>
<td>2006</td>
<td>–</td>
<td>0.6</td>
</tr>
<tr>
<td>Zheng et al. (2012)</td>
<td>2009</td>
<td>38.3</td>
<td>0.4</td>
</tr>
<tr>
<td>This work</td>
<td>2006</td>
<td>52.3</td>
<td>0.6</td>
</tr>
</tbody>
</table>

* Including emissions from industry, power generation, transportation, and residential activities.
Table 2. PRD carbonaceous aerosol emissions: bottom-up estimates used to drive our model and top-down estimates obtained from this study.

<table>
<thead>
<tr>
<th>Source sectors</th>
<th>EC (GgC yr(^{-1}))</th>
<th>OC (GgC yr(^{-1}))</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Bottom-up</td>
<td>Top-down</td>
</tr>
<tr>
<td>Anthropogenic</td>
<td>38.9</td>
<td>52.3 ± 8.0 (±25%)</td>
</tr>
<tr>
<td>Industry</td>
<td>7.2(^a)</td>
<td>39.7 ± 7.4 (±37%)</td>
</tr>
<tr>
<td>Power generation</td>
<td>1.1(^a)</td>
<td>39.7 ± 7.4 (±37%)</td>
</tr>
<tr>
<td>Residential sources</td>
<td>4.9(^a)</td>
<td>12.6 ± 3.1 (±48%)</td>
</tr>
<tr>
<td>Transportation</td>
<td>25.7(^a)</td>
<td>12.6 ± 3.1 (±48%)</td>
</tr>
<tr>
<td>Biomass burning</td>
<td>0.6(^b)</td>
<td>0.6 ± 0.8 (±133%)</td>
</tr>
<tr>
<td>Total</td>
<td>39.5 (±161%)(^c)</td>
<td>52.9 ± 8.0 (±30%)</td>
</tr>
</tbody>
</table>

\(^a\) From Zheng et al. (2012) for the inner PRD and from Zhang et al. (2009) for Hong Kong and Macau. The uncertainties (represented as 95% confidence intervals, CI) for residential emissions in the Zheng et al. (2012) inventory were 2.0–10.9 Gg C yr\(^{-1}\) for EC and 5.5–24.8 Gg C yr\(^{-1}\) for OC. The uncertainties for emissions from other sectors in the Zheng et al. (2012) inventory were not quantified. The 95% CI for the Zhang et al. (2009) inventory are ±208% for EC and ±258% for OC.

\(^b\) Emissions and uncertainties were from M. He et al. (2011) for the inner PRD and from Streets et al. (2003b) for Hong Kong and Macau.

\(^c\) The overall uncertainties (95% CI) are shown in parentheses, calculated from sectorial emission uncertainties for residential sources (Zheng et al., 2012), for biomass burning (M. He et al., 2011), and for other anthropogenic sectors (Lei et al., 2011).

\(^d\) Uncertainties are represented as ± standard deviations. The corresponding 95% CI (calculated assuming normality) are shown in parentheses.
Table 3. VOC precursors of secondary organic carbon in the PRD area in 2006.

<table>
<thead>
<tr>
<th>VOC precursor</th>
<th>Emissions over the PRD area (Ggyr⁻¹)</th>
<th>Simulated annual mean SOC concentrations (µgC m⁻³) (^d)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Anthropogenic(^a)</td>
<td>Biomass burning(^b)</td>
</tr>
<tr>
<td>Isoprene</td>
<td>0.6</td>
<td>0.1</td>
</tr>
<tr>
<td>Monoterpenes</td>
<td>1.7</td>
<td>0.1</td>
</tr>
<tr>
<td>Alkanes</td>
<td>113</td>
<td>0.1</td>
</tr>
<tr>
<td>Ethylene</td>
<td>68.2</td>
<td>4.3</td>
</tr>
<tr>
<td>Xylenes</td>
<td>77.0</td>
<td>0.2</td>
</tr>
<tr>
<td>Toluene</td>
<td>51.7</td>
<td>0.1</td>
</tr>
<tr>
<td>Acetone</td>
<td>8.8</td>
<td>2.1</td>
</tr>
<tr>
<td>Benzene</td>
<td>18.7</td>
<td>0.5</td>
</tr>
<tr>
<td>Acetylene</td>
<td>16.6</td>
<td>0.1</td>
</tr>
<tr>
<td>Propene</td>
<td>6.5</td>
<td>0.7</td>
</tr>
<tr>
<td>Glyoxal</td>
<td>4.8</td>
<td>1.1</td>
</tr>
<tr>
<td>Glycolaldehyde</td>
<td>3.6</td>
<td>0.8</td>
</tr>
<tr>
<td>Methylyglyoxal</td>
<td>3.2</td>
<td>0.7</td>
</tr>
<tr>
<td>Hydroxyacetone</td>
<td>3.0</td>
<td>0.7</td>
</tr>
<tr>
<td>Cresols</td>
<td>&lt; 0.1</td>
<td>–</td>
</tr>
<tr>
<td>Phenol</td>
<td>–</td>
<td>&lt; 0.1</td>
</tr>
<tr>
<td>Sesquiterpenes</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td>378</td>
<td>11</td>
</tr>
</tbody>
</table>

\(^a\) Includes industry, power generation, transportation, and residential sources from the inventories developed by Zheng et al. (2012) and Zhang et al. (2009).

\(^b\) From the inventories developed by M. He et al. (2011) and Streets et al. (2003b). Biomass burning emissions of glyoxal, methylyglyoxal, glycolaldehyde, and hydroxyacetone were scaled from biomass burning CO emissions.

\(^c\) Calculated in CMAQ using the MEGAN algorithm (Guenther et al., 2006).

\(^d\) From CMAQ simulation using the top-down EC and OC emission estimates described in the text. SOC concentrations are averaged over urban and rural sites of available seasonal observations.
Table 4. List of the PRD area carbonaceous aerosol surface concentration measurements used in this study.

<table>
<thead>
<tr>
<th>Location</th>
<th>Aerosol size sampled</th>
<th>Filter accumulation period/sampling frequency</th>
<th>Observation period</th>
<th>Representative season</th>
<th>Observed seasonal mean surface concentrations [µgC m(^{-3})]</th>
<th>Method for EC and OC analysis</th>
<th>Method for determining SOC</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Huangpu (HP, 113.45° N, 23.10° E)</td>
<td>PM(_{10})</td>
<td>24 h/every 6 days</td>
<td>01.2002–02.2002</td>
<td>Winter</td>
<td>12.1 (10.5) 32.6 (28.5) 15.0 (15.8)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{10})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>10.5 (15.8)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Liwan (LW, 113.25° E, 23.12° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002, 12.2002, 03.2003</td>
<td>Annual</td>
<td>4.4 (17.6) 17.6 (10.1)</td>
<td>TOT</td>
<td>EC tracer</td>
<td>[3]</td>
</tr>
<tr>
<td>Longgui (LG, 113.28° E, 23.27° N)</td>
<td>PM(_{10})</td>
<td>24 h/every 6 days</td>
<td>03.2003</td>
<td>Winter</td>
<td>13.3 (8.8) 38.4 (24.7) 19.1 (14.1)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>8.8 (14.1)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[2]</td>
</tr>
<tr>
<td>Lo wu (LH, 114.12° E, 22.55° N)</td>
<td>PM(_{10})</td>
<td>24 h/every 3 days</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>7.3 (10.4) 16.4 (4.4)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>5.0 (4.4)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Shenzhen graduate school</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002, 12.2002, 03.2003</td>
<td>Annual</td>
<td>3.9 (11.1) 4.5 (4.5)</td>
<td>TOT</td>
<td>EC tracer</td>
<td>[3]</td>
</tr>
<tr>
<td>Peking University (PKU, 113.93° E, 22.53° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>03.2003</td>
<td>Autumn</td>
<td>6.2 (11.3) 4.5 (4.5)</td>
<td>–</td>
<td>AMS/PMF</td>
<td>[4]</td>
</tr>
<tr>
<td>South China Institute (SCI, 113.15° E, 23.07° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>01.2008–11.2008</td>
<td>Winter</td>
<td>7.9 (24.2) 15.9 (5.9)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[5]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>01.2008–11.2008</td>
<td>Winter</td>
<td>7.9 (24.2) 15.9 (5.9)</td>
<td>AMS/PMF</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sun Yat-sen University (SU, 113.28° E, 23.08° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>01.2002–02.2002</td>
<td>Winter</td>
<td>8.1 (23.3) 11.6 (10.8)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Winter</td>
<td>5.9 (17.8) 10.8 (9.8)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>05.2005–07.2007</td>
<td>Summer</td>
<td>6.5 (16.4) 8.1 (5.6)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>05.2005–07.2007</td>
<td>Spring</td>
<td>4.3 (8.7) 3.4 (3.4)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>05.2005–07.2007</td>
<td>Summer</td>
<td>3.4 (6.5) 3.5 (3.5)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>05.2005–07.2007</td>
<td>Autumn</td>
<td>5.5 (11.0) 5.9 (5.9)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>01.2002–02.2002</td>
<td>Winter</td>
<td>6.0 (14.5) 5.8 (5.8)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Winter</td>
<td>2.5 (6.9) 3.8 (3.8)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>08.2006–08.2007</td>
<td>Winter</td>
<td>4.8 (9.4) 3.2 (3.2)</td>
<td>TOR</td>
<td>EC tracer</td>
<td>[6]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>08.2006–08.2007</td>
<td>Summer</td>
<td>3.4 (6.5) 2.7 (2.7)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>08.2006–08.2007</td>
<td>Summer</td>
<td>3.0 (5.7) 1.2 (1.2)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>08.2006–08.2007</td>
<td>Summer</td>
<td>5.6 (8.9) 2.0 (2.0)</td>
<td>EC tracer</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

a TOR: thermal optical reflectance; TOT: thermal optical transmission; TMO: thermal manganese dioxide oxidation.
SOC is calculated as OC – EC (POC/EC), where the (POC/EC) value is obtained from the seasonal or annual minimum observed OC/EC ratio.
Longgui was reported as a background site in Cao et al. (2003), but its EC and OC concentrations are very high and show obvious influence from local sources. We categorized it as an urban site for this study.
Use AMS to collect organic matter (OM) and aethalometer to collect BC.
The AMS measurements can be analyzed using the PMF technique to identify four OA components: HOA, BBOA, SV-OOA, and LV-OOA. We used the total oxygenated OA (SV-OOA + LV-OOA) as a surrogate for SOA.
These sites were reported as background sites in Cao et al. (2003, 2004), Ho et al. (2003), Hagler et al. (2006), and Gnauk et al. (2008). We categorize them as rural sites for this study, because they reflect the baseline pollution levels in the PRD area.
Use isoprene, monoterpenes, β-caryophyllene, and toluene as SOA tracers to estimate SOA concentration in PM\(_{2.5}\) samples.
Table 4. Continued.

<table>
<thead>
<tr>
<th>Location</th>
<th>Aerosol size sampled</th>
<th>Filter accumulation period/sampling frequency</th>
<th>Observation period</th>
<th>Representative season</th>
<th>Observed seasonal mean surface concentrations [µg C m(^{-2})]</th>
<th>Method for EC and OC analysis (^a)</th>
<th>Method for determining SOC</th>
<th>Reference (^b)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hong Kong</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Baptist University (BU, 114.20° E, 22.32° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>01.2002–02.2002</td>
<td>Winter</td>
<td>3.4 7.6 2.7</td>
<td>TOR EC tracer</td>
<td>TOT EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{10})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>3.9 6.7 2.0</td>
<td>TOR EC tracer</td>
<td>TOT EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td>Central and western (CW, 114.13° E, 22.28° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002,12.2002,03.2003,06.2003</td>
<td>Annual</td>
<td>1.9 6.6 3.4</td>
<td>TOT EC tracer</td>
<td>TOT EC tracer</td>
<td>[3]</td>
</tr>
<tr>
<td>Hong Tsui (HT, 114.25° E, 22.22° N)</td>
<td>PM(_{10})</td>
<td>24 h/every 6 days</td>
<td>01.2002–02.2002</td>
<td>Winter</td>
<td>3.0 9.1 4.8</td>
<td>TOR EC tracer</td>
<td>TOT EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{10})</td>
<td>24 h/daily</td>
<td>06.2002–07.2002</td>
<td>Summer</td>
<td>1.1 4.1 2.8</td>
<td>TOR EC tracer</td>
<td>TOT EC tracer</td>
<td>[1]</td>
</tr>
<tr>
<td></td>
<td>PM(_{10})</td>
<td>24 h/daily</td>
<td>11.2000–02.2001</td>
<td>Winter</td>
<td>1.5 5.6 –</td>
<td>TMO –</td>
<td>TOT EC tracer</td>
<td>[7]</td>
</tr>
<tr>
<td>Hong Kong University of Science and Technology (UST, 114.25° E, 22.33° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 2 days</td>
<td>07.2006</td>
<td>Summer</td>
<td>1.0 4.7 4.1</td>
<td>TOT SOA tracer</td>
<td>TOT SOA tracer</td>
<td>[8]</td>
</tr>
<tr>
<td>Tap Mun (TM, 114.35° E, 22.47° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002,12.2002,03.2003,06.2003</td>
<td>Annual</td>
<td>0.8 4.9 3.5</td>
<td>TOT EC tracer</td>
<td>TOT EC tracer</td>
<td>[3]</td>
</tr>
<tr>
<td>Tsuen Wan (TW, 114.10° E, 22.37° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 2 days</td>
<td>07.2006</td>
<td>Summer</td>
<td>3.5 8.1 5.7</td>
<td>TOT SOA tracer</td>
<td>TOT SOA tracer</td>
<td>[8]</td>
</tr>
<tr>
<td>Tung Chung (TC, 113.93° E, 22.28° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002,12.2002,03.2003,06.2003</td>
<td>Annual</td>
<td>2.0 6.3 2.9</td>
<td>TOT EC tracer</td>
<td>TOT EC tracer</td>
<td>[3]</td>
</tr>
<tr>
<td></td>
<td>PM(_{2.5})</td>
<td>24 h/every 2 days</td>
<td>07.2006</td>
<td>Summer</td>
<td>1.6 5.5 5.0</td>
<td>TOT SOA tracer</td>
<td>TOT SOA tracer</td>
<td>[8]</td>
</tr>
<tr>
<td>Yuen Long (YL, 114.02° E, 22.43° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 2 days</td>
<td>07.2006</td>
<td>Summer</td>
<td>2.4 7.8 4.9</td>
<td>TOT SOA tracer</td>
<td>TOT SOA tracer</td>
<td>[8]</td>
</tr>
<tr>
<td>Rural</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Back Garden (BG, 113.03° E, 23.49° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/daily</td>
<td>07.2006</td>
<td>Summer</td>
<td>3.3 5.7 2.0</td>
<td>TOT EC tracer</td>
<td>TOT EC tracer</td>
<td>[9]</td>
</tr>
<tr>
<td>Xinken (XK, 113.58° E, 22.65° N)</td>
<td>PM(_{10})</td>
<td>20 h/daily</td>
<td>10.2004</td>
<td>Autumn</td>
<td>6.1 7.0 2.4</td>
<td>– EC tracer</td>
<td>TOT SOA tracer</td>
<td>[11]</td>
</tr>
<tr>
<td>Zhongshan (ZS, 113.41° E, 22.52° N)</td>
<td>PM(_{2.5})</td>
<td>24 h/every 6 days</td>
<td>10.2002,12.2002,03.2003,06.2003</td>
<td>Annual</td>
<td>2.5 10.5 6.3</td>
<td>TOT EC tracer</td>
<td>TOT EC tracer</td>
<td>[3]</td>
</tr>
</tbody>
</table>

\(^a\) TOR: thermal optical reflectance; TOT: thermal optical transmission; TMO: thermal manganese dioxide oxidation.


SOC is calculated as OC\(^-\)EC \(\times\) (POC/EC), where the (POC/EC) value is obtained from the seasonal or annual minimum observed OC/EC ratio.

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Use isoprene, monoterpenes, \(\beta\)-caryophyllene, and toluene as SOA tracers to estimate SOA concentration in PM\(_{2.5}\) samples.
**Table 5.** Numbers of seasonal carbonaceous aerosol measurement sites in the PRD area used in this study.

<table>
<thead>
<tr>
<th></th>
<th>Spring</th>
<th>Summer</th>
<th>Autumn</th>
<th>Winter</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Urban*</td>
<td>6</td>
<td>10</td>
<td>6</td>
<td>10</td>
<td>32</td>
</tr>
<tr>
<td>Rural</td>
<td>2</td>
<td>3</td>
<td>4</td>
<td>2</td>
<td>11</td>
</tr>
<tr>
<td>Total</td>
<td>8</td>
<td>13</td>
<td>10</td>
<td>12</td>
<td>43</td>
</tr>
</tbody>
</table>

* All locations in Hong Kong are averaged to represent a single urban site, HK.
Figure 1. Locations of surface carbonaceous aerosol measurements used in this study, including 15 urban locations (circles) and 8 rural locations (triangles). Details of measurements at each location are presented in Table 4.
Figure 2. Annual emissions of EC in the PRD area in 2006 from the bottom-up inventories (Zheng et al., 2012; M. He et al., 2011; Zhang et al., 2009; and Streets et al., 2003b) used to drive our simulation. PRD annual emission totals are shown inset in red. For the sake of clarity, emissions outside the PRD area are not shown.
Figure 3. Annual emissions of OC in the PRD area in 2006 from the bottom-up inventories (Zheng et al., 2012; M. He et al., 2011; Zhang et al., 2009; and Streets et al., 2003b) used to drive our simulation. PRD annual emission totals are shown inset in red.
Figure 4. Annual emissions of anthropogenic xylenes, biomass burning glyoxal, and biogenic isoprene over the PRD area in 2006. PRD annual emission totals are shown inset in red.
Figure 5. Seasonal mean surface EC concentrations and simulated source contributions at 15 urban and rural sites over the PRD area in 2006. The left panel shows seasonal mean observed surface EC concentrations at urban (circles) and rural (triangles) sites over the PRD area. Underlain are simulated seasonal mean EC concentrations in (a) spring, (b) summer, (c) autumn and (d) winter in 2006 using top-down emission estimates. Simulated PRD seasonal mean concentrations are shown inset in red. The right panel shows the simulated seasonal mean surface EC concentrations contributed by anthropogenic transportation sources (blue), anthropogenic non-transportation (industrial, power generation and residential) sources (cyan), biomass burning sources (yellow), and outside of the PRD area (orange) in (e) spring, (f) summer, (g) autumn and (h) winter in 2006 using the bottom-up and the top-down emission inventories. Also shown are observed EC concentrations (green squares), as well as simulated EC concentrations with bottom-up (pink diamonds) and top-down emissions (yellow circles).
**Figure 6.** Seasonal mean surface OC concentrations and simulated source contributions at 15 urban and rural sites over the PRD area in 2006. The left panel shows seasonal mean observed surface OC concentrations at urban (circles) and rural (triangles) sites over the PRD area. Underlain are simulated seasonal mean OC concentrations in (a) spring, (b) summer, (c) autumn and (d) winter in 2006 using top-down emission estimates. Simulated PRD seasonal mean concentrations are shown inset in red. The right panel shows the simulated seasonal mean surface OC concentrations contributed by anthropogenic transportation sources (blue), anthropogenic non-transportation (industrial, power generation and residential) sources (cyan), biomass burning sources (yellow), outside of the PRD area (orange), and secondary sources (green) in (e) spring, (f) summer, (g) autumn and (h) winter in 2006 using the bottom-up and the top-down emission inventories. Also shown are observed OC concentrations (green squares), as well as simulated OC concentrations with bottom-up (pink diamonds) and top-down emissions (yellow circles).
Figure 7. Simulated vs. observed seasonal surface (a) EC, (b) OC, and (c) POC concentrations at urban (circles) and rural (triangles) sites. Also shown are the reduced-major axis regression lines (solid lines): model results using bottom-up emission inventories (black), and model results using top-down emission estimates (red). The regression slopes ($S$), correlation coefficients ($r$), and numbers of data points ($n$) are shown inset. The grey dashed lines indicate the $1:1$ lines.
Figure 8. Top-down estimates of carbonaceous aerosol emissions in the PRD area in 2006. The annual emissions summed over the PRD area are shown inset in red.
Figure 9. Simulated components of surface annual mean OC using top-down emission estimates: (a) anthropogenic primary OC; (b) biomass burning primary OC; (c) non-PRD primary OC (transport from outside of the PRD area); (d) semi-volatile and non-volatile oligomeric SOC from biogenic and anthropogenic VOC precursors; (e) SOC from irreversible uptake of dicarbonyls by aqueous-phase particles (mainly in clouds); and (f) total simulated OC. Simulated annual mean surface concentrations for each component are shown inset in red.