

1     **Intra-annual variations of regional aerosol optical depth, vertical distribution, and**  
2     **particle types from multiple satellite and ground-based observational datasets**

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## 11 **Abstract**

12           The climatic and health effects of aerosols are strongly dependent on the intra-annual  
13 variations in their loading and properties. While the seasonal variations of regional aerosol optical  
14 depth (AOD) have been extensively studied, understanding the temporal variations in aerosol  
15 vertical distribution and particle types is also important for an accurate estimate of aerosol climatic  
16 effects. In this paper, we combine the observations from four satellite-borne sensors and ground-  
17 based AOD and fine particle (PM<sub>2.5</sub>) measurements to investigate the seasonal variations of aerosol  
18 column loading, vertical distribution, and particle types over three populous regions: the Eastern  
19 United States (EUS), Western Europe (WEU), and Eastern and Central China (ECC). In all three  
20 regions, column AOD, as well as AOD at height above 800 m, peaks in summer/spring probably  
21 due to accelerated formation of secondary aerosols and hygroscopic growth. In contrast, AOD  
22 below 800 m peaks in winter over WEU and ECC regions because more aerosols are confined to  
23 lower heights due to the weaker vertical mixing. In the EUS region, AOD below 800 m shows two  
24 maximums, one in summer and the other in winter. The temporal trends in low-level AOD are  
25 consistent with those in surface PM<sub>2.5</sub> concentrations. AOD due to fine particles (< 0.7 μm diameter)  
26 is much larger in spring/summer than in winter over all three regions. The coarse mode AOD (>  
27 1.4 μm diameter), however, generally shows small variability except that a peak occurs in spring  
28 in the ECC region due to the prevalence of airborne dust during this season. When aerosols are  
29 classified according to sources, the dominant type is associated with anthropogenic air pollution,  
30 which has a similar seasonal pattern as total AOD. Dust and sea-spray aerosols in the WEU region  
31 peak in summer and winter, respectively, but do not show an obvious seasonal pattern in the EUS  
32 region. Smoke aerosols, as well as absorbing aerosols, present an obvious unimodal distribution  
33 with a maximum occurring in summer over the EUS and WEU regions, whereas they follow a

34 bimodal distribution with peaks in August and March (due to crop residue burning) over the ECC  
35 region.

36

## 37 **1 Introduction**

38 Aerosols have adverse effects on human health (Lelieveld et al., 2015) and play a key role  
39 in Earth's climate through aerosol-radiation interactions (McCormick and Ludwig, 1967) and  
40 aerosol-cloud interactions (Twomey, 1977; Albrecht, 1989; Garrett and Zhao, 2006). Compared  
41 with long-lived climate forcers such as CO<sub>2</sub>, aerosols have relatively short lifetimes and hence  
42 large spatiotemporal variability (Unger et al., 2008; Shindell et al., 2009). Therefore, the climatic  
43 and health effects of aerosols are not only induced by inter-annual concentration changes, but also  
44 strongly depend on their intra-annual variability.

45 Aerosol optical depth (AOD) has been widely used to represent the column aerosol loading  
46 and to assess the aerosol impacts on radiation, clouds, and precipitation (Ma et al., 2014; Niu and  
47 Li, 2012; Zhao et al., 2018b; Song et al., 2017). However, the wide ranges of particle optical  
48 properties and size distribution mean that even for the same AOD, different aerosol types have  
49 different effects on not only the magnitude, but also the sign, of aerosol radiative forcing (IPCC,  
50 2013; Gu et al., 2006; Garrett et al., 2004). IPCC (2013) estimates that the historical global mean  
51 direct radiative forcings due to sulfate, organic carbon (OC), black carbon (BC), and mineral dust  
52 are  $-0.40$ ,  $-0.19$ ,  $+0.36$ , and  $-0.10$  W m<sup>-2</sup>, respectively. Furthermore, absorbing and non-  
53 absorbing aerosols have been found to have very different impacts on the surface radiative cooling  
54 effects (Yang et al., 2016) and the development of convective clouds (Massie et al., 2016;  
55 Ramanathan et al., 2005; Rosenfeld et al., 2008). Besides aerosol type, the aerosol vertical  
56 distribution influences its mass concentration within the planetary boundary layer (PBL) (Zheng

57 et al., 2017) and the vertical profile of heating rate (Johnson et al., 2008; Guan et al., 2010; Zhang  
58 et al., 2013), which subsequently modifies the atmospheric stability and convective strength  
59 (Ramanathan et al., 2007), with potential changes in cloud properties (Johnson et al., 2004).  
60 Understanding aerosol variability as a function of height is also important because the indirect  
61 effect of aerosols is mainly dependent on those mixed with the clouds (Zhao et al., 2018c).  
62 Meanwhile, the health impacts of aerosols are only associated with those present near the surface,  
63 where they are inhaled. For these reasons, systematic analyses of the intra-annual variations of  
64 aerosol vertical distribution and particle types, in addition to total column AOD, are necessary to  
65 improve our understanding of aerosol climatic and health effects.

66 Numerous studies have investigated the seasonal variations of AOD at global and regional  
67 scales using satellite observations (e.g., Kim et al., 2007; Song et al., 2009; Mehta et al., 2016;  
68 Mao et al., 2014). By comparison, most previous studies of the temporal variations of aerosol  
69 vertical distributions and aerosol types have been confined to only a few sites due to coverage  
70 limitations associated with reliance on ground-based instruments (e.g., Liu et al., 2012; Matthias  
71 et al., 2004). Despite continuous advancement of remote sensing technology and emergence of  
72 new spaceborne sensors, only limited number of studies have utilized satellite observations to  
73 examine the seasonal variations of aerosol profiles and/or types at regional or larger scales (Huang  
74 et al., 2013; Kahn and Gaitley, 2015; Yu et al., 2010; Li et al., 2016). Huang et al. (2013) analyzed  
75 the seasonal variations of aerosol extinction profile and type distribution using 5-year observations  
76 from the Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO). Kahn  
77 and Gaitley (2015) examined the spatiotemporal variations of aerosol types retrieved by the Multi-  
78 angle Imaging SpectroRadiometer (MISR). Different satellite-borne sensors, such as MISR,  
79 CALIPSO, and Moderate resolution Imaging Spectroradiometer (MODIS), employ different

80 principles of measurement and retrieval, and therefore provide different sensitivities to column  
81 AOD, aerosol types, and vertical profiles. Therefore, integration of data from multiple satellites  
82 and ground-based observational networks makes it possible to deepen our understanding of the  
83 intra-annual variations of aerosol loadings, profiles, and types.

84 In this study, we investigate the seasonal variations of aerosol column loading, vertical  
85 distribution, and particle types using multiple satellite and ground-based observational datasets  
86 covering the period from 2007 to 2016. The purpose is to assess the consistency among various  
87 datasets and provide a comprehensive characterization of aerosol properties in polluted regions to  
88 facilitate future studies of aerosol climate effects and local air quality issues. The data are from  
89 MISR, MODIS, CALIPSO, Aerosol Robotic Network (AERONET), and surface PM<sub>2.5</sub> monitors.  
90 Following our previous study (Zhao et al., 2017), we selected three populous regions which have  
91 experienced substantial anthropogenic pollution (Wang et al., 2017; Wang et al., 2014) and have  
92 received considerable attention in other climate studies: the Eastern United States (EUS; 29°-45°  
93 N, 70°-98° W), Western Europe (WEU; 37°-59° N, 10° W-17° E), and Eastern and Central China  
94 (ECC; 21°-41° N, 102°-122° E). The geographical boundaries of these regions are shown in Fig. 1.

## 95 **2 Data and Methods**

### 96 2.1 Satellite data

97 We obtain retrievals of total column AOD as well as AOD for various height ranges and  
98 aerosol types during 2007-2016 from MISR (flying on the Terra satellite), MODIS (Terra and  
99 Aqua), and the Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP) on CALIPSO. The  
100 aerosol retrievals from MISR and MODIS are only available for clear-sky conditions in the  
101 daytime. CALIPSO provides retrievals during both day and night, but only clear-sky daytime  
102 profiles are used in order to be consistent with the products from MISR and MODIS.

103 MISR observes the Earth with moderately high spatial resolution (275 m to 1.1 km) at 9  
104 along-track viewing angles in each of 4 visible/near-infrared spectral bands, which enables the  
105 partitioning of AOD by particle type over both land and ocean, in addition to retrieval of total  
106 AOD (Kahn and Gaitley, 2015; Kahn et al., 2001). Its observations provide near-global coverage  
107 every 9 days (Diner et al., 1998). We make use of the Level 3 daily global aerosol product  
108 (MIL3DAE) version F15\_0031, which is generated at a spatial resolution of  $0.5^\circ \times 0.5^\circ$  based on  
109 the Level 2 aerosol product V22. The variables used in the analysis are total AOD at 555 nm as  
110 well as AODs for six aerosol components, namely small ( $< 0.7 \mu\text{m}$  diameter), medium (0.7-1.4  
111  $\mu\text{m}$  diameter), large ( $> 1.4 \mu\text{m}$  diameter), spherical, non-spherical, and absorbing. Based on  
112 comparison with ground-based AERONET measurements, the errors in MISR Level 2 AOD data  
113 are on the order of  $\pm 0.05$  or  $\pm(0.20 \times \text{AOD})$ , whichever is larger (Kahn et al., 2005; Kahn et al.,  
114 2010). In addition, retrieval of MISR aerosol type information from individual retrievals is  
115 considered to be reliable when  $\text{AOD} > 0.15$ , and has diminished sensitivity at smaller AOD (Kahn  
116 and Gaitley, 2015; Kahn et al., 2010). In this study we use only monthly mean values, for which  
117 the uncertainties in aerosol types are expected to be smaller than those for individual retrievals.  
118 Note that we did not do a relative humidity (RH) correction to AOD retrievals from MISR as well  
119 as other sensors. The seasonal variations of AOD represent a combined effect of variations in  
120 aerosol abundance, vertical distribution, chemical constituents, and meteorological conditions.

121 The MODIS sensors onboard the Terra and Aqua satellites observe the Earth with multiple  
122 wavelength bands over a 2330 km swath (King et al., 2003), which provides near-daily global  
123 coverage. In this study we obtain column AOD data at 550 nm with a  $1^\circ \times 1^\circ$  resolution from the  
124 Level 3 daily atmosphere products Collection 6 (MOD08 and MYD08 for the Terra and Aqua  
125 platforms, respectively). Comparison studies with AERONET have estimated the accuracy of

126 Level 2 AOD retrievals to be about  $\pm(0.05 + 0.15 \times \text{AOD})$  over land and  $\pm(0.03 + 0.05 \times \text{AOD})$   
127 over ocean (Levy et al., 2010; Remer et al., 2005). For both MISR and MODIS data, we calculate  
128 regional mean AOD by averaging valid AOD values over all grids within the three target regions.

129 CALIOP is a dual-wavelength polarization lidar on the CALIPSO satellite, and is designed  
130 to acquire vertical profiles of aerosols and clouds at 532 and 1064 nm wavelengths (Winker et al.,  
131 2007). CALIPSO flies in formation with Aqua, and all three satellites employed in this paper fly  
132 in orbits having 16-day repeat cycles. In addition to vertical extinction profiles, CALIPSO  
133 categorizes an aerosol layer as one of seven types based on a number of parameters including  
134 altitude, location, surface type, volume depolarization ratio, and integrated attenuated backscatter  
135 (Omar et al., 2009). The seven aerosol types are dust, smoke, clean continental, polluted  
136 continental, polluted dust, clean marine, and dusty marine. For most profiles, this aerosol  
137 classification is consistent with that derived from AERONET inversion data (Mielonen et al.,  
138 2009). In this study, we adopt the Level 2 aerosol profile product (05kmAPro, V4.10), which has  
139 an along-track horizontal resolution of 5 km and a vertical resolution of 60 m or 180 m, depending  
140 on whether the aerosol height is below or above 20.2 km altitude. We do not use the CALIOP  
141 Level 3 product because it is difficult to collocate with AERONET observations (see Section 2.2)  
142 due to its coarse resolution ( $2^\circ \times 5^\circ$ ). For each clear-sky profile, we calculate the column AOD at  
143 532 nm by vertically integrating extinction coefficients of the features that are identified as  
144 “aerosols” and have valid quality control (QC) flags, i.e.,  $-100 \leq \text{cloud aerosol discrimination}$   
145 (CAD) score  $\leq -20$ , extinction QC = 0/1, and extinction coefficient uncertainty  $< 99.9$  (Huang et  
146 al., 2013). In addition, we employ two quality filters used in generating the Level 3 product in  
147 order to eliminate features that probably suffer from surface contamination, i.e., near-surface  
148 features with large negative extinction coefficients and contaminated features beneath the surface-

149 attached opaque layer (NASA CALIPSO team, 2011). Following the same method, we also bin  
150 the 532 nm AODs into various height ranges, i.e., 0-200 m, 200-500 m, 500-800m, 800-1200 m,  
151 1200-2000 m, and > 2000 m above ground level (AGL). Finally, we derive monthly mean AODs  
152 by averaging all clear-sky aerosol profiles within each month over the three target regions.  
153 Although aerosol extinction coefficients with heights below 200 m AGL are considered to be  
154 uncertain despite the application of quality filters (NASA CALIPSO team, 2011), we include them  
155 for completeness but exercise with caution when interpreting variations of AODs below 200 m. It  
156 should be noted that CALIPSO AOD is reported at a different wavelength (532 nm) from those  
157 used in the MISR and MODIS products (555 nm and 550 nm, respectively); this slight wavelength  
158 difference is not expected to affect our conclusions regarding AOD seasonal variations.

## 159 2.2 AERONET and surface PM<sub>2.5</sub> data

160 We use AOD observations from AERONET to compare with the AOD seasonal variations  
161 derived from satellite datasets. AERONET sunphotometers directly measure AOD at seven  
162 wavelengths (approximately 340, 380, 440, 500, 675, 870, and 1020 nm) with an estimated  
163 uncertainty of 0.01-0.02 (Holben et al., 2001; Eck et al., 1999), which is much smaller than the  
164 uncertainties associated with satellite measurements (Kahn et al., 2010; Levy et al., 2010; Schuster  
165 et al., 2012). Therefore, we consider AERONET as “ground truth” for AOD temporal variations.  
166 We adopt the AERONET Level 2 Version 2.0 direct-sun measurements of spectral AODs, which  
167 are subsequently interpolated to 550 nm using a second-order polynomial fit to  $\ln(\text{AOD})$  vs.  
168  $\ln(\text{wavelength})$  as recommended by Eck et al. (1999). A fundamental difference between satellite  
169 and AERONET AOD observations is that a satellite acquires data at a single overpass time (or  
170 spread over 7 minutes for MISR’s nine views) and over an extended spatial area in the case of  
171 MISR and MODIS, whereas AERONET obtains a time series of point data at each surface station.



172 To match coincident measurements, the AERONET AOD retrievals for each site are averaged  
173 within a 2 h window centered on the satellite overpass times (about 10:30 for MISR and  
174 MODIS/Terra, and 13:30 for MODIS/Aqua and CALIPSO, depending on site location), and  
175 compared with the satellite AOD retrievals in a  $1^\circ \times 1^\circ$  grid box (consistent with the grids used in  
176 the MODIS Level 3 products) that contains the corresponding AERONET site. Only those days  
177 for which a satellite overpasses an AERONET site are used in the comparisons. Since AOD  
178 variation has a large spatial correlation length of 40-400 km (Anderson et al., 2003), spatial  
179 averaging over a  $1^\circ \times 1^\circ$  grid should not bias the seasonal variations of AOD but has the benefit  
180 of increase the number of data points with valid AOD retrievals that are used in the comparisons.  
181 To assure data quality, only the AERONET sites that span at least 5 years with at least 10 months  
182 of valid data in each year are included in the comparison. After screening, 28, 54, and 13 sites are  
183 used in our analysis of the EUS, WEU, and ECC regions.

184 To provide additional information on the seasonal variations of satellite-observed aerosol  
185 loadings near the surface, we obtain surface  $PM_{2.5}$  concentrations from several observational  
186 networks over the three target regions. Hourly  $PM_{2.5}$  concentrations for 225 sites over the EUS  
187 region are achieved from the Air Quality System (AQS), which is a large observational database  
188 containing ambient air pollution data collected by the United States Environmental Protection  
189 Agency (USEPA), as well as state, local, and tribal air pollution control agencies in the United  
190 States (USEPA, 2015). For the ECC region, we obtain hourly  $PM_{2.5}$  concentrations from the  
191 Ministry of Environmental Protection of China (MEP, <http://datacenter.mep.gov.cn/>), which  
192 provides continuous measurements at 496 sites located in 74 major cities in China. Hourly/daily  
193  $PM_{2.5}$  concentrations for 52 sites over the WEU region are taken from the European Monitoring  
194 and Evaluation Programme (EMEP). Similar to the processing of AERONET data, we only include

195 sites whose data span  $\geq 5$  years with  $\geq 10$  months of data in each year, except in the case of the  
196 ECC region where at least 2 years' data are required because the PM<sub>2.5</sub> concentrations have been  
197 only publicly available since January 2013.

### 198 **3 Results and Discussion**

#### 199 3.1 Seasonal variations of column AOD

200 Figure 2 illustrates the monthly variations in column AOD observed by MISR,  
201 MODIS/Terra, MODIS/Aqua, and CALIPSO during 2007-2016 in the three target regions. All  
202 satellite-borne sensors show that AOD in the EUS region is the highest in summer and lowest in  
203 winter, though CALIPSO reports a noticeably smaller difference between the summer and winter  
204 extrema compared with the other three satellite instruments. For the WEU and ECC regions, MISR,  
205 MODIS/Terra, and MODIS/Aqua also reveal consistent seasonal patterns in which AOD peaks in  
206 spring and/or summer and reaches its lowest valley in winter. CALIPSO, however, shows little  
207 intra-annual variation in AOD, with small peaks occurring in spring and fall.

208 As described in Section 2.1, MODIS provides near-daily global coverage but MISR and  
209 CALIPSO do not. As a result, the monthly mean AOD from different sensors is calculated based  
210 on different sets of days, which might lead to uncertainties in the estimation of monthly mean  
211 AOD (Colarco et al., 2014; Wang and Zhao, 2017). To rule out the impact of spatio-temporal  
212 sampling on seasonal variation patterns, we design two sensitivity cases: a “MODIS/Terra\_match  
213 MISR” case in which the monthly mean AOD of MODIS/Terra is calculated using only the days  
214 when MISR overpasses, and a “MODIS/Aqua\_match CALIPSO” case in which the monthly mean  
215 AOD of MODIS/Aqua is calculated using only the overpassing days of CALIPSO. The results are  
216 illustrated in Fig. 2. In all three regions, the monthly mean AODs are slightly different for  
217 “MODIS/Terra” and “MODIS/Terra\_match MISR”, but the seasonal variation patterns are largely

218 the same. The same results are found for “MODIS/Aqua” and “MODIS/Aqua\_match CALIPSO”.  
219 As such, we conclude that sampling has little effect on the AOD seasonal variation patterns  
220 reported in this study. In fact, this conclusion is compatible with the findings of Colarco et al.  
221 (2014). Colarco et al. (2014) revealed that the spatial sampling artifacts were significant for fine  
222 aggregation grid (e.g., 0.5°), but they are reduced at coarse grid scales (e.g., 10°). In this study, we  
223 use only the mean AOD over three large regions (about 20°×20°) across 10 years, therefore the  
224 sampling artifacts are expected to be even smaller. Even though, we acknowledge that the  
225 inconsistent spatiotemporal sampling of different retrieval products (due to different swath width  
226 and mixing of Level 2 and Level 3 products) adds to the uncertainty in monthly AOD estimation.  
227 A more direct comparison at the measurement/retrieval level merits further in-depth study.

228         In view of the substantial differences between CALIPSO and the other three sensors, we  
229 compare satellite retrieved AOD seasonal variations with point-based ground measurements from  
230 AERONET (Fig. 3). As in other studies, AERONET data are treated as “ground truth” for column  
231 AOD due to its smaller uncertainty compared with satellite data (Kahn et al., 2010; Levy et al.,  
232 2010; Schuster et al., 2012; Fan et al., 2018). Figure 3 shows that, in all three regions, the AOD  
233 seasonal variations measured by AERONET are similar to those retrieved by MISR, MODIS/Terra,  
234 and MODIS/Aqua, but are quite different from CALIPSO data. Reasons for the different seasonal  
235 patterns between CALIPSO and other sensors will be discussed in Section 3.2. Considering the  
236 high accuracy of AERONET, we conclude that AOD peaks in summer/spring and dips in winter.  
237 An important reason for the higher AOD in summer is that the stronger radiation and higher  
238 temperature accelerate the formation of secondary aerosols (Timonen et al., 2014), including  
239 sulfate, nitrate, ammonium, and secondary organic aerosol (SOA). SOA is produced by photo-  
240 oxidation of volatile organic compounds (VOCs) and intermediate volatility organic compounds

241 (IVOCs), as well as the chemical aging of primary organic aerosol (Zhao et al., 2016). Another  
242 reason is that more abundant water vapor in summer favors the hygroscopic growth of aerosols  
243 (Liu et al., 2012; Zheng et al., 2017). The different patterns of long range transport as a function  
244 of season is also partly responsible for the seasonable variation of AOD (Tian et al., 2017; Yang  
245 et al., 2018; Garrett et al., 2010).

246 While relative patterns of AOD seasonal variations from observations of MISR,  
247 MODIS/Terra, and MODIS/Aqua are similar to each other and to those of AERONET, the  
248 magnitude of AOD observed by these sensors shows considerable discrepancies. In all three  
249 regions, the AOD retrieved from MODIS is larger than that from MISR, consistent with the results  
250 of previous studies (de Meij et al., 2012; Zhao et al., 2017; Chin et al., 2014; Kang et al., 2016; Qi  
251 et al., 2013). This is most likely due to differences in observing strategy, retrieval algorithms, and  
252 spatio-temporal sampling (Kahn et al., 2009). The MISR-retrieved AOD agrees well with the  
253 AERONET observations in EUS and WEU regions. In the ECC region, however, MISR  
254 underestimates the AERONET AOD, probably because there is less signal from the surface at  
255 higher AOD, which creates ambiguity that can result in the algorithm assigning too much of the  
256 top-of-atmosphere radiance to the surface (i.e., a higher surface albedo), thereby underestimating  
257 the AOD (Kahn et al., 2010). The MODIS/Terra and MODIS/Aqua overestimate the AERONET  
258 AOD to some extent in all three regions. The overestimation was also reported in two previous  
259 studies (de Meij et al., 2012; Ruiz-Arias et al., 2013) using the level 3 MODIS products (Collection  
260 5 or 5.1). We show a relatively larger overestimation than that reported by de Meij et al. (2012)  
261 and Ruiz-Arias et al. (2013), partly because we used the AERONET AOD averaged within a 2 h  
262 window centered on the satellite overpass times while the two previous studies used the  
263 daily/monthly mean AERONET AOD in the comparisons. The daily mean AOD observed by

264 AERONET is about 10% larger than the value during the satellite overpass times (Li et al., 2013).  
265 The reasons for the discrepancy between MODIS and AERONET are yet to be thoroughly  
266 investigated.

### 267 3.2 Seasonal variations of aerosol loadings as a function of height

268 In addition to column AOD, the climatic effects of aerosols are also strongly dependent on  
269 their vertical distribution. To explore intra-annual variations in aerosol vertical profile, Fig. 4  
270 presents CALIPSO-observed monthly variations of AOD as a function of height in the three target  
271 regions. A striking pattern is that the AOD seasonal variations are dramatically different at lower  
272 and upper heights. Over the WEU and ECC regions, AODs of the vertical layers below 800 m  
273 AGL generally peak in winter, while those above 800 m AGL peak in summer/spring. As a result,  
274 the CALIPSO-observed column AOD for these two regions presents a rather uniform seasonal  
275 pattern. For the EUS region, the maximum AOD above 800 m AGL also occurs in summer;  
276 however, AOD below 800 m AGL shows two peaks, one in summer and the other in winter. The  
277 integration of various layers thus yields a nearly unimodal distribution with maximum occurring  
278 in summer.

279 To provide an independent evaluation of the CALIPSO-observed AOD variations at lower  
280 heights, we examine the seasonal variations of near-surface  $PM_{2.5}$  concentrations at hundreds of  
281 surface monitor locations within the three target regions (Figure 5). The aerosol extinction  
282 coefficient, and hence AOD at lower heights is affected by not only the particle mass  
283 concentrations, but also aerosol type (absorbing vs. nonabsorbing aerosols, coarse-mode vs. fine-  
284 mode aerosols) and meteorological parameters such as RH, wind speed and direction, and  
285 planetary boundary layer height (Zheng et al., 2017). Nevertheless, previous studies have reported  
286 fairly good correlations between extinction coefficient/low-level AOD and  $PM_{2.5}$  concentrations

287 (Cheng et al., 2013; Zheng et al., 2017). For this reason, it is reasonable to qualitatively compare  
288 the seasonal variation patterns of near-surface PM<sub>2.5</sub> concentrations and low-level AOD. We  
289 calculate monthly mean PM<sub>2.5</sub> concentrations using only the days when CALIPSO overpasses an  
290 observational site to enable a better comparison. Figure 5 shows that, over the ECC and WEU  
291 regions, surface PM<sub>2.5</sub> concentrations are largest in winter and smallest in summer. In the EUS  
292 region, the maximum PM<sub>2.5</sub> concentration occurs in summer and a second maximum occurs in  
293 winter. These trends are generally consistent with the seasonal variations of AOD at low heights,  
294 implying that CALIPSO data can generally capture the seasonal changes in low-level aerosol  
295 abundance.

296         The aerosol vertical distribution is an important factor in reconciling CALIPSO and other  
297 sensors with regard to AOD seasonal variations. MISR, MODIS, and AERONET all measure  
298 column-integrated AOD using spectroradiometers, whereas CALIOP is an active lidar which  
299 estimates vertically-resolved AOD based on vertical profiles of attenuated backscatter. By  
300 comparing CALIPSO with the Atmospheric Radiation Measurement (ARM) program's ground-  
301 based Raman lidars, Thorsen et al. (2017) showed that CALIPSO does not detect all relatively  
302 significant aerosols due to insufficient detection sensitivity and tends to miss optically thin aerosol  
303 layers. Consequently, the fraction of aerosols detected in the upper levels (> 800 m AGL) is much  
304 smaller than that in the lower levels (< 800 m AGL) because the upper-level aerosols are often  
305 optically thin. As a result, the CALIPSO-observed AOD seasonal variations are significantly  
306 weighted toward lower heights. Note that the aerosols with heights below 200 m AGL are  
307 frequently undetected because of surface contamination (Kim et al., 2017; NASA CALIPSO team,  
308 2011), but this does not alter the key feature that the AOD is weighted toward lower heights. Over  
309 WEU and ECC regions, the unimodal AOD distributions with a summer peak at higher levels are

310 largely counteracted by the opposite seasonal variations at lower levels, resulting in rather uniform  
311 seasonal variations of column AOD. For the EUS regions, due to the bimodal AOD distribution at  
312 lower heights, the summer peak in column AOD variations remain but the difference between peak  
313 and valley is smaller than implied by the observations of MISR/MODIS/AERONET. In this sense,  
314 although the integrated CALIPSO column AOD does not agree well with AERONET, it does  
315 provide valuable information with respect to seasonal variations of aerosols within a specific  
316 height range. This is because the detection fraction of aerosols does not vary significantly with  
317 season at a given height due to relatively small variability of optical thickness. Specifically, the  
318 seasonal mean AOD within a specific height range differs by at most 3 times as a function of  
319 season (Fig. 4), while it decreases by about 2 orders of magnitude with the increase of height (Kim  
320 et al., 2017; Thorsen et al., 2017). Besides the seasonal variations, the difference in the magnitude  
321 of AOD between CALIPSO and other sensors are also largely explained by the undetected aerosol  
322 layers by CALIPSO (Kim et al., 2017; Thorsen et al., 2017) as well as the assumed lidar ratios in  
323 CALIPSO retrievals (Ma et al., 2013).

324         Why are the AOD seasonal variations different between the lower and upper levels? The  
325 atmosphere in winter is generally more stable and vertical mixing is weaker, therefore more  
326 aerosols, particularly primary aerosols, are confined to lower heights, resulting in the peak of low-  
327 level AOD in winter (Guo et al., 2016; Liu et al., 2012; Zheng et al., 2017). At higher levels, the  
328 maximum AOD in summer can be explained by two reasons: (1) more aerosols, especially primary  
329 aerosols, are transported to the upper levels in summer due to stronger vertical mixing (Guo et al.,  
330 2016; Liu et al., 2012; Zheng et al., 2017), and (2) secondary aerosol formation is more rapid in  
331 summer because of stronger radiation and higher temperature, and much of the secondary aerosols  
332 are produced in the upper levels (de Reus et al., 2000; Minguillon et al., 2015; Heald et al., 2005).

333 In addition, the seasonal variations of AOD at different vertical levels may also be influenced by  
334 the variations of water vapor amount which affects the hygroscopic growth (Liu et al., 2012; Zheng  
335 et al., 2017) as well as the seasonal patterns of inter-regional transport of aerosols (Tian et al.,  
336 2017; Yang et al., 2018; Garrett et al., 2010).

### 337 3.3 Seasonal variations of aerosol types

338 Besides column AOD and vertical profiles, another factor influencing aerosol climate  
339 impact is aerosol type (i.e., partitioning by size and chemical composition). The MISR and  
340 CALIPSO products classify aerosols based on distinct principles of measurement and retrieval  
341 algorithms. Analysis of the two datasets in combination can potentially lead to a deeper  
342 understanding of the factors driving temporal variations of aerosol type.

343 Figures 6 illustrates the seasonal variations of type-specific AODs retrieved by MISR.  
344 MISR distributes AODs into three size ranges, i.e., small ( $< 0.7 \mu\text{m}$  diameter), medium ( $0.7\text{-}1.4$   
345  $\mu\text{m}$  diameter), and large ( $> 1.4 \mu\text{m}$  diameter). The ambient aerosols are comprised of primary  
346 aerosols (dust, sea-spray aerosols, and primary anthropogenic aerosols) and secondary aerosols  
347 (sulfate, nitrate, ammonium, and SOA). Among these constituents, dust and sea-spray aerosols are  
348 predominantly coarse particles and secondary aerosols are dominated by very fine particles, while  
349 primary anthropogenic aerosols span a large size range, leading to a mean size intermediate  
350 between dust/sea-spray and secondary constituents (Seinfeld and Pandis, 2006). Fig. 6 indicates  
351 that the small-size AOD is much larger in spring/summer than in winter over all regions, primarily  
352 due to accelerated secondary aerosol formation and enhanced hygroscopic growth (see Section  
353 3.1). In contrast, large-size AOD generally shows rather uniform distributions, except for the ECC  
354 region where a peak occurs in late winter/early spring. AOD of primary anthropogenic aerosols  
355 are less influenced by seasonal effects than secondary aerosols, which partly accounts for the rather



356 uniform distributions of large-size AOD. Additionally, the seasonal variations of large-size AOD  
357 are also affected by dust and sea-spray aerosols, as discussed below.

358 In contrast to MISR's partitioning of aerosol type by size and absorption, the CALIPSO-  
359 retrieved aerosol types are characterized by emission source (Fig. 7). As discussed in Section 3.2,  
360 relative variability in CALIPSO-derived AOD at different height ranges appears to be more  
361 reliable than integrated column AOD, therefore we show aerosol types below and above 800 m  
362 separately in Fig. 7. Particles associated with anthropogenic air pollution (polluted continental and  
363 polluted dust) comprise the dominant type in all three regions. The seasonal variation patterns of  
364 polluted continental/dust are in accordance with those of the total AOD. Specifically, at higher  
365 levels, the maximum AOD of polluted continental/dust aerosols occurs in spring/summer in all  
366 regions. At lower levels, however, the maximum occurs in winter (plus a second maximum in  
367 summer in EUS).

368 With regard to dust and clean marine (sea-spray) aerosols, the AOD in the EUS region does  
369 not show an obvious seasonal pattern. In the WEU region, AOD of dust aerosols peaks in summer,  
370 consistent with previous surface-based observational studies which show that dust events in  
371 Europe predominantly occur during summer due to transport from the Sahara region (Stafoggia et  
372 al., 2016). The AOD of dust is primarily located above 800 m, supporting the conclusion that dust  
373 aerosols in WEU mainly originate from long range transport. Since the dust AOD is subject to a  
374 large inter-annual variability (denoted by the large error bars in Fig. 7), we use the Student's t-test  
375 to demonstrate the statistical significance of the seasonal variations. The dust AOD in summer is  
376 statistically larger than that in any other season at the 0.05 level, indicating the robustness of the  
377 peak in summer. Contrary to dust, the AOD of sea-spray aerosols in WEU is much higher in winter  
378 than in summer, probably because winter is the relative windy season with large low pressure

379 systems over the Atlantic Ocean and the North Sea (Manders et al., 2009). The offset of the  
380 opposite variation trends in dust and sea-spray aerosols partly accounts for the rather uniform  
381 distributions of large-size AOD in WEU (see Fig. 6). Over the ECC region, sea-spray aerosols  
382 make a negligible contribution to total AOD. The dust AOD is much larger in spring than in any  
383 other season (significant at the 0.05 level), which is tied to the outburst of springtime Gobi desert  
384 dust storms (China Meteorological Administration, 2012). The high dust AOD explains the peak  
385 in large-size AOD in spring over the ECC region (see Fig. 6).

386         Smoke aerosols are predominantly located above 800 m in all regions. Over the EUS and  
387 WEU regions, smoke aerosols present a unimodal distribution with maximum occurring in summer.  
388 The differences between smoke AOD in summer and the other three seasons are all statistically  
389 significant at the 0.05 level, except for the difference between summer and spring over the WEU  
390 region, which is statistically significant at the 0.10 level. In the ECC region, the smoke AOD  
391 follows a bimodal distribution with peaks occurring in March and August and valleys occurring in  
392 May and December. The differences between either of the peak months and either of the valley  
393 months are statistically significant at the 0.05 level. MISR's independent retrieval of absorbing  
394 AOD (Fig. 6) presents a highly similar seasonal pattern (statistically significant at the 0.05 level)  
395 as the CALIPSO smoke AOD. In fact, smoke and absorbing aerosols are closely correlated with  
396 each other, since smoke consists of a much larger fraction of absorbing aerosols (Dubovik et al.,  
397 2002), such as BC and light-absorbing organic aerosol (Kirchstetter and Thatcher, 2012), as  
398 compared to other aerosol types. Besides, the MISR absorbing AOD and CALIPSO smoke AOD  
399 are also consistent in the order of magnitude. The variability of MISR absorbing AOD (shown in  
400 the right Y-axis of Fig. 6) is about 0.002-0.005, while the variability of smoke AOD from  
401 CALIPSO is about 0.01-0.03. The smoke AOD includes the contributions of both the absorbing

402 and scattering portions. The MISR absorbing AOD, which is calculated using total AOD  $\times$  (1 –  
403 single scattering albedo), represents only the absorbing portion but includes contributions from  
404 aerosol types other than smoke (Bull et al., 2011). Considering that the single scattering albedo of  
405 smoke is about 0.80-0.94 (Dubovik et al., 2002), we are able to reconcile the magnitude of MISR  
406 absorbing AOD and CALIPSO smoke AOD. For the preceding reasons, the seasonal patterns of  
407 smoke and absorbing aerosols act as a cross-validation and strengthen the reliability of the  
408 observed trends. Over the EUS and WEU regions, the largest smoke AOD in summer could be  
409 explained by the highest emissions from forest and grassland fires (van der Werf et al., 2017).  
410 Over the ECC region, an additional peak occurs in March because agricultural residue burning  
411 makes a substantial contribution to total smoke emissions (van der Werf et al., 2017), and such  
412 burning takes place more frequently in March due to burning of crop residues left on the fields  
413 from the previous growing season (Shon, 2015).

#### 414 **4 Conclusions and implications**

415 This study investigated the seasonal variations of aerosol column loading, vertical  
416 distribution, and particle types using multiple satellite and ground-based observational datasets  
417 during 2007-2016 over EUS, WEU, and ECC regions. Retrievals from MISR and MODIS reveal  
418 that column AOD in all three regions peaks in spring/summer and reaches its low in winter, which  
419 is consistent with observations from AERONET. This seasonal pattern is probably explained by  
420 accelerated formation of secondary aerosols in spring/summer due to stronger insolation and  
421 higher temperature. In contrast, CALIPSO shows a much weaker seasonal variability in column  
422 AOD, probably because CALIPSO-retrieved AOD is weighted toward lower heights since some  
423 thin aerosol layers in high levels are undetected due to insufficient detection sensitivity. Despite  
424 the discrepancy in integrated column AOD, CALIPSO does provide valuable information with

425 respect to intra-annual variations of AOD as a function of height. Over the WEU and ECC regions,  
426 AODs of the vertical layers below 800 m generally peak in winter, while those above 800 m mostly  
427 peak in summer. For the EUS region, the maximum AOD above 800 m also occurs in summer;  
428 however, AOD below 800 m shows two peaks, one in summer and the other in winter. The seasonal  
429 variations of AOD at low heights are consistent with seasonal patterns of measured surface PM<sub>2.5</sub>  
430 concentrations.

431         When aerosols are binned into different size ranges, the small-size AOD is much larger in  
432 spring/summer than in winter over all three regions. Large-size AOD generally shows rather  
433 uniform distributions, except for the ECC region where a peak occurs in spring, consistent with  
434 the largest dust AOD in this season. When aerosols are classified according to sources, the aerosols  
435 associated with anthropogenic air pollution (as well as mixtures of anthropogenic pollution and  
436 dust) are the dominant type in all three regions. AOD of polluted aerosols has a similar seasonal  
437 pattern as total AOD. Dust and clean marine aerosols in the WEU region peak in summer and  
438 winter, respectively, whereas they do not show an obvious seasonal pattern in the EUS region.  
439 Smoke aerosols, which CALIPSO indicates are predominantly located at heights above 800 m,  
440 present an obvious unimodal distribution with maximum occurring in summer over EUS and WEU  
441 regions, and a bimodal distribution with peaks in August and March over the ECC region. This  
442 pattern is in good agreement with the seasonal variations of absorbing AOD derived from MISR.

443         The combination of multiple satellite and ground-based observations facilitate a systematic  
444 and deeper understanding of the seasonal variations of aerosols, particularly their vertical and type  
445 distribution. Comparison of multiple measurement and retrieval methodologies enables reducing  
446 the uncertainties in the estimation of aerosol direct effects by providing improved information  
447 about aerosol vertical and type distributions, which significantly affect the aerosol-induced

448 scattering and absorption of radiation. More importantly, the intra-annual variations of vertical  
449 distributions and types of aerosols are important for understanding their impact on atmospheric  
450 dynamics, cloud fields, and precipitation production (Ramanathan et al., 2005; Massie et al., 2016;  
451 Zhao et al., 2018a; Wang et al., 2013). Finally, the data and variation patterns presented in this  
452 study can be used to evaluate and improve model simulations, with the ultimate goal of improving  
453 model assessment of the climatic and health effects of aerosols.

454

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461

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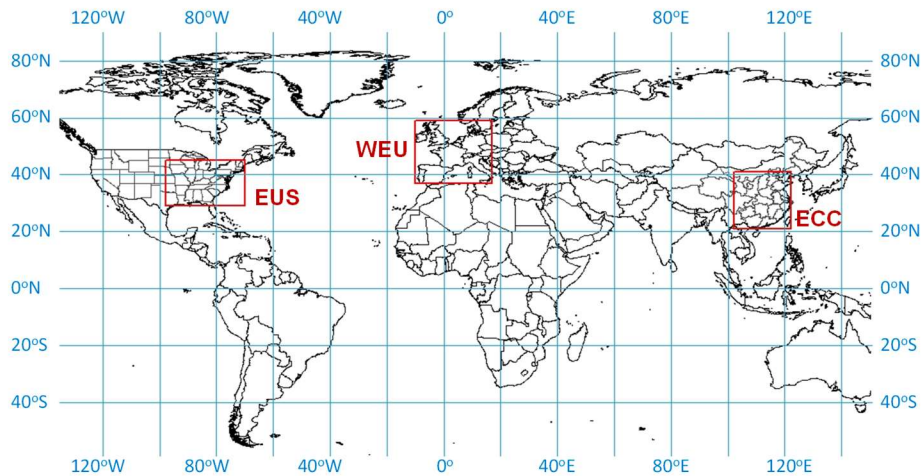
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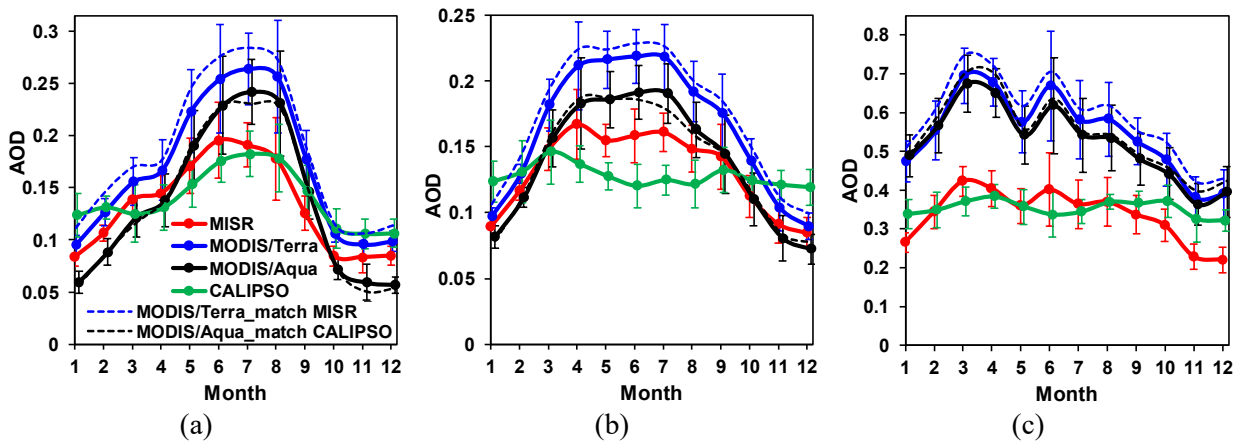
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721 **Figures**



722 **Figure 1.** Target regions for this study: the Eastern United States (EUS), Western Europe (WEU),  
723 and Eastern and Central China (ECC).  
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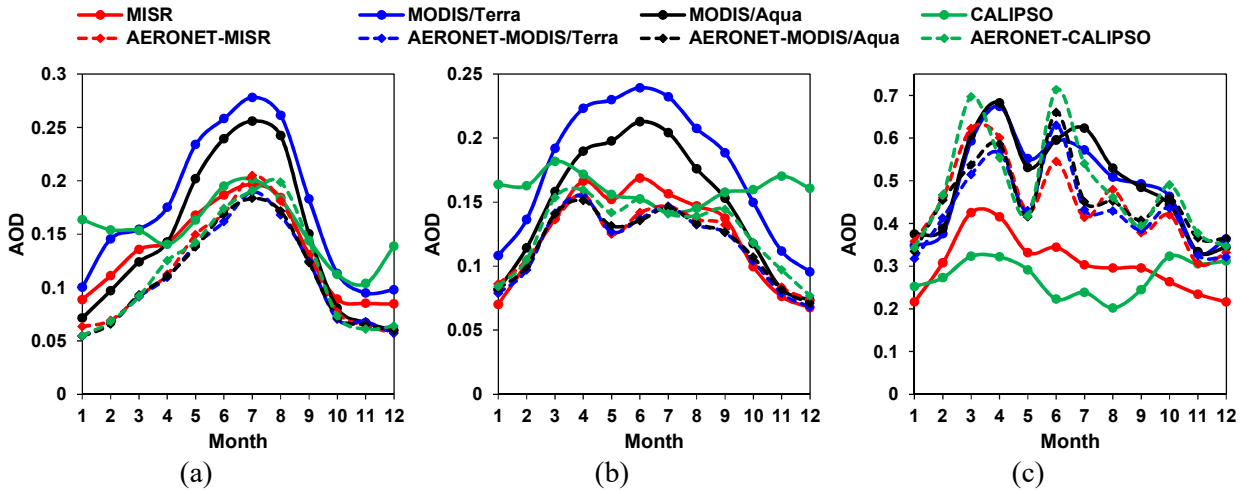
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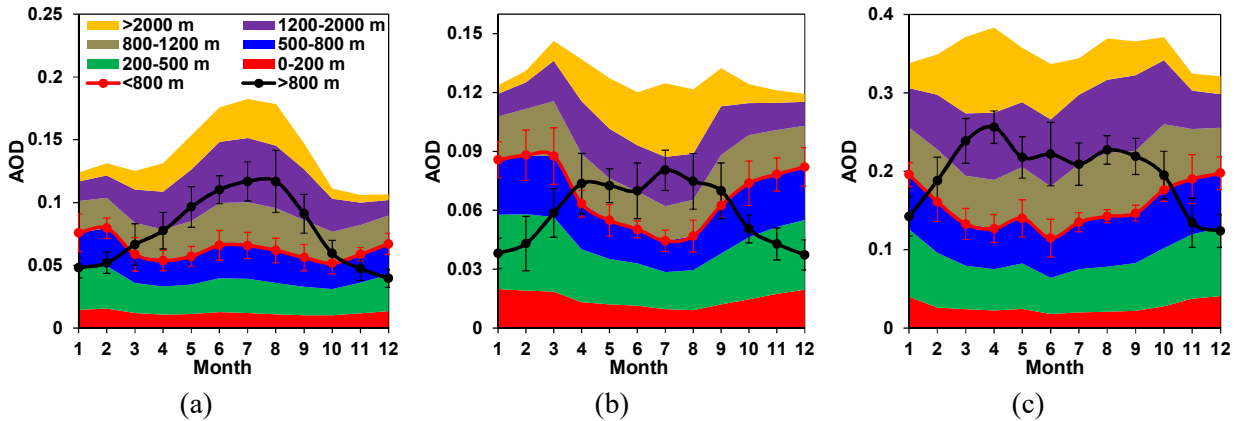
726 **Figure 2.** Monthly mean AOD observed by MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO  
727 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. For CALIPSO, only clear-sky daytime  
728 profiles are averaged in order to be consistent with the MISR and MODIS products.  
729 “MODIS/Terra\_match MISR” is a sensitivity case in which the monthly mean AOD of MODIS/Terra  
730 is calculated using only the days when MISR overpasses, and “MODIS/Aqua\_match CALIPSO”  
731 is a case in which the monthly mean AOD of MODIS/Aqua is calculated using only the

732 overpassing days of CALIPSO. The error bars denote the standard deviation of the monthly mean  
 733 AOD values obtained over all years. Note the different scales on the y-axes of the plots.

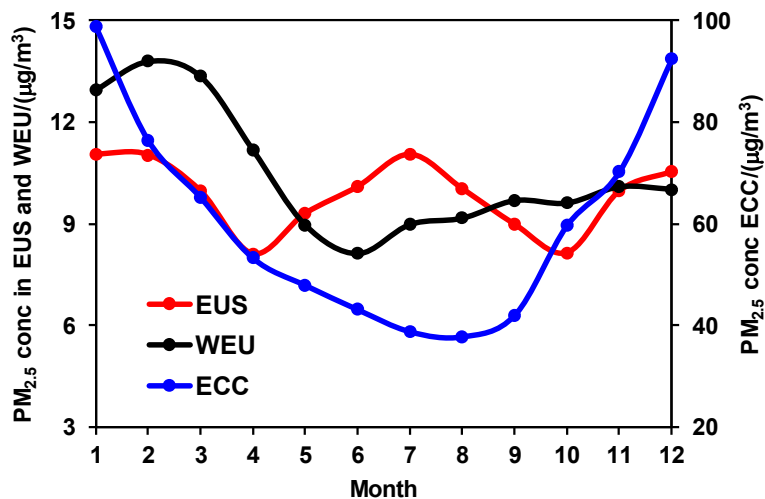
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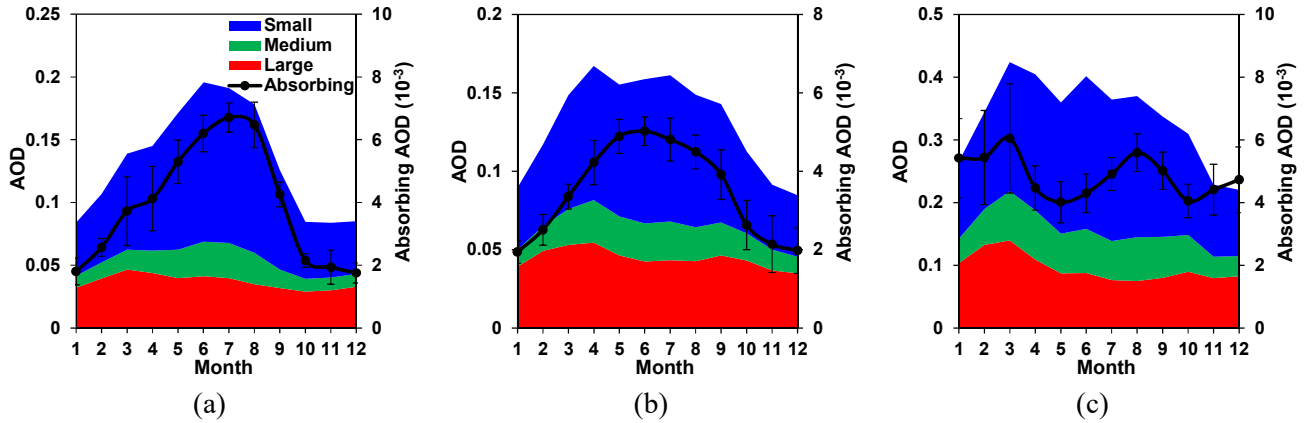
735 **Figure 3.** Monthly mean AOD observed by satellites and AERONET averaged across the  
 736 AERONET sites during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. The observations from  
 737 MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO are averaged over  $1^{\circ} \times 1^{\circ}$  grid boxes containing  
 738 the AERONET sites. The AERONET data are averaged within a 2 h window centered on satellite  
 739 overpass times. The numbers of AERONET sites included in analysis are 28, 54, and 13, in the  
 740 EUS, WEU, and ECC regions, respectively. Since the four sensors overpass a site in different  
 741 days and different times of day, we separately calculate the AERONET data matched to each  
 742 sensor (denoted by “AERONET-xxx”). The AERONET curves matched to different sensors are  
 743 close in EUS and WEU, partly because there are plenty of sites in these two regions, and the  
 744 discrepancy due to the sampling issue is therefore smoothed out. In contrast, there are only 13  
 745 AERONET sites in ECC, so there exists larger discrepancy between the AERONET data matched  
 746 to different sensors. Note the different scales on the y-axes of the plots.



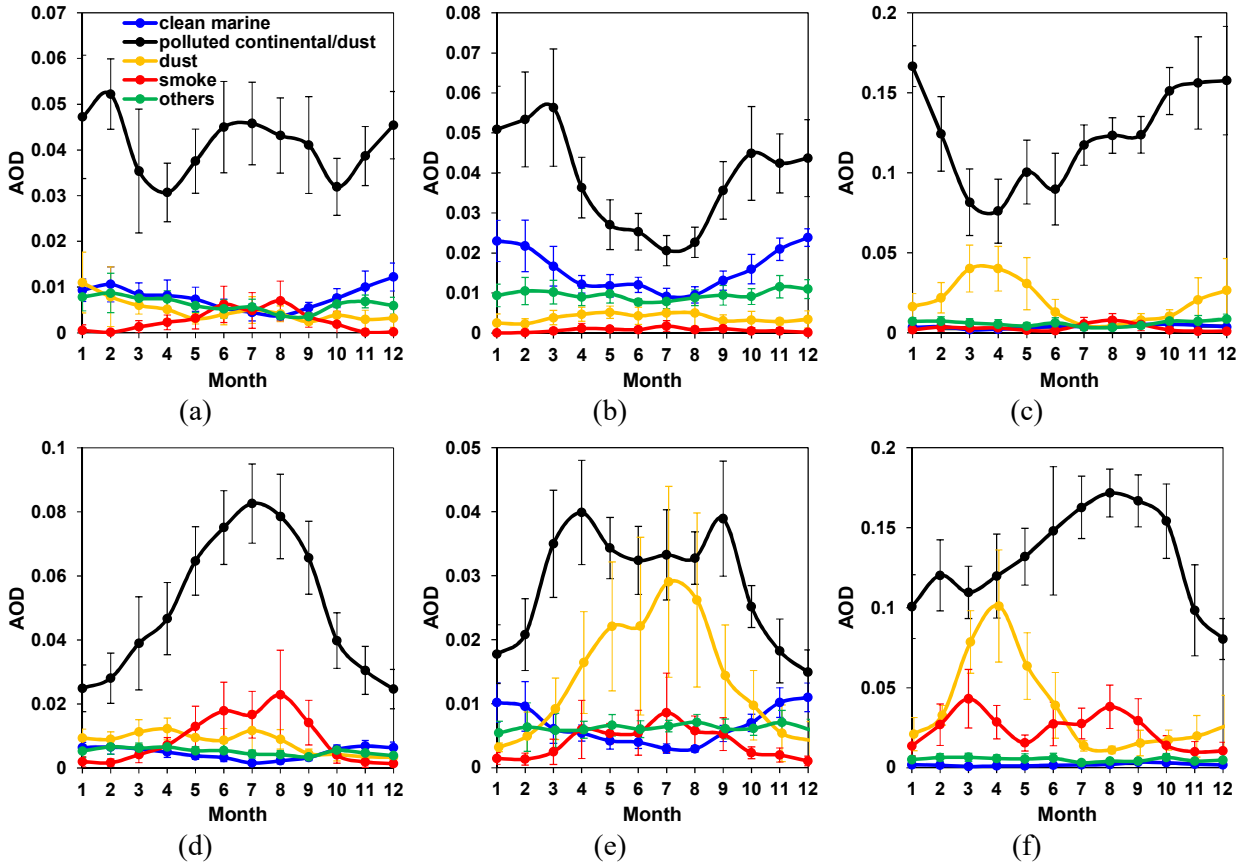
747 **Figure 4.** Monthly mean AOD as a function of height above ground level observed by CALIPSO  
 748 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. Only clear-sky daytime profiles are averaged  
 749 in order to be consistent with the products of MISR and MODIS. The range of AOD within a  
 750 particular height range is depicted by the colored stacks. The integrated AODs for heights below  
 751 and above 800 m are shown as solid lines, for which the error bars are defined in the same way  
 752 as in Fig. 2. Note the different scales on the y-axes of the plots.



753 **Figure 5.** Monthly mean surface  $PM_{2.5}$  concentrations during 2007-2016 in three target regions.  
 754 The numbers of observational sites included in averaging are 225, 52, and 496, in the EUS,  
 755 WEU, and ECC regions. Note the different scales on the y-axes for EUS/WEU and ECC.  
 756



757 **Figure 6.** Monthly mean AOD of different aerosol types observed by MISR during 2007-2016 in  
 758 (a) EUS, (b) WEU, and (c) ECC. The size-resolved AODs are depicted by the colored stacks (left  
 759 Y-axis); the integration of the three size ranges yields total column AOD, as represented by the  
 760 upper edge of the blue color. The AOD of absorbing aerosols is shown as solid lines (right Y-axis),  
 761 for which the error bars are defined in the same way as in Fig. 2. Note the different scales on the  
 762 y-axes of the plots.



763 **Figure 7.** Monthly mean AOD of different aerosol types (a-c) below 800 m and (d-f) above 800 m  
 764 observed by CALIPSO during 2007-2016 in (a, d) EUS, (b, e) WEU, and (c, f) ECC. Only clear-  
 765 sky daytime profiles are used in the averaging to be consistent with the products of MISR and  
 766 MODIS. The definition of error bars is the same as in Fig. 2. Note the different scales on the y-  
 767 axes of the plots.

768 **Table 1.** Summary of the seasonal variations of the total, height-specific, and type-specific AOD

	<b>EUS</b>	<b>WEU</b>	<b>ECC</b>
Total column AOD	Peak in summer	Peak in summer/late spring	Peak in summer/spring
AOD > 800 m AGL	Peak in summer	Peak in summer/late spring	Peak in summer/spring
AOD < 800 m AGL	Two peaks in winter and summer	Peak in winter	Peak in winter
Small-size	Peak in summer	Peak in summer/late spring	Peak in summer/spring
Medium-size	Peak in summer	Peak in summer/late spring	Peak in summer/spring
Large-size	Rather uniform	Rather uniform	Peak in spring
Absorbing	Peak in summer	Peak in summer/late spring	Two peaks in Mar and Aug
Polluted continental/dust	Similar to height-specific total AOD	Similar to height-specific total AOD	Similar to height-specific total AOD
Dust	No obvious seasonal pattern	Peak in summer	Peak in spring
Clean marine	No obvious seasonal pattern	Peak in winter	Negligible amount
Smoke	Peak in summer	Peak in summer/late spring	Two peaks in Mar and Aug

769