

Reviewer 1:

Aerosol radiative effect is a hot topic in the science community, which is dependent on the aerosol optical properties, size distribution, aerosol types, and their vertical distribution. While many studies have examined the seasonal and diurnal variations of aerosols, few studies have examined the vertical distributions of aerosol amount and types, which can strongly affect the aerosol radiative effect and corresponding thermal impacts on the profiles of temperature. Using multi-satellite observations along with the surface observations of aerosols, this study examines the seasonal and diurnal variations of aerosol column loading, vertical distribution, and particle types over three populous regions: the Eastern United States (EUS), Western Europe (WEU), and Eastern and Central China (ECC). Interesting statistical characteristics about the dominant aerosol types and vertical distributions have been obtained. The paper is also organized and written well.

Response: We thank the reviewer for the valuable comments. We have followed these comments in revising the manuscript. Point-to-point responses are given below.

Detailed Comments:

Line 45-48: I am not sure if the climate effects of CO<sub>2</sub> also depends on its intra-annual variability, particularly considering the higher outgoing longwave radiation from surface in summer than in winter.

Response: We no longer mention the climate effects of CO<sub>2</sub> in the revised manuscript. The revised sentence is as follows:

Therefore, the climatic and health effects of aerosols are not only induced by inter-annual concentration changes, but also strongly depend on their intra-annual variability. (Line 41-43)

Line 49-54: Not only the scattering and absorption properties, but also the size distribution and vertical distribution of aerosols can also cause problems. Zheng et al. (2017, ACP) have shown that the vertical distribution of AOD could have strong impacts on the aerosol concentration (mass concentration) within PBL, which directly affect cloud properties and then change cloud radiative forcing. Garrett et al. (2004, GRL) showed that nucleation mode aerosols and accumulation mode aerosols have very different scattering radiative effects.

Response: Following this comment, we have described the impact of size distribution and vertical distribution and cited these two references. The revised text is as follows:

However, the wide ranges of particle optical properties and size distribution mean that even for the same AOD, different aerosol components have different effects on not only the magnitude, but also the sign, of aerosol radiative forcing (IPCC, 2013; Gu et al., 2006; Garrett et al., 2004). (Line 46-49)

Besides aerosol type, the aerosol vertical distribution influences its mass concentration within the planetary boundary layer (PBL) (Zheng et al., 2017) and the vertical profile of heating rate (Johnson et al., 2008; Guan et al., 2010; Zhang et al., 2013), which subsequently modifies the atmospheric stability and convective strength (Ramanathan et al., 2007), with potential changes in cloud properties (Johnson et al., 2004). (Line 54-58)

Line 54-58: Besides to the TOA radiative balance, convective clouds development, absorptive and non-absorptive aerosols have different impacts on the surface radiative cooling effects, as shown by Yang et al. (2016, JGR) which demonstrated that more absorptive aerosol can cause more surface cooling effects.

Response: We have mentioned the different impacts of absorptive and non-absorptive aerosols on surface radiative cooling effects in the revised manuscript:

Furthermore, absorbing and non-absorbing aerosols have been found to have very different impacts on the surface radiative cooling effects (Yang et al., 2016) and the development of convective clouds (Massie et al., 2016; Ramanathan et al., 2005; Rosenfeld et al., 2008). (Line 51-54)

Line 59-63: The vertical distributions are also important for aerosol-cloud interaction since only the aerosols near cloud bases have strong interaction with cloud properties. Zhao et al. (2018, Earth and Space Science) showed Twomey effect using in-situ aircraft observations in East China instead of anti-Twomey effect found using column aerosols based on satellite observations.

Response: Following the reviewer's comments, we have added this point in the revised manuscript:

Understanding aerosol variability as a function of height is also important because the indirect effect of aerosols is mainly dependent on those mixed with the clouds (Zhao et al., 2018b). (Line 59-60)

Line 101-103: You may indicate the data observation time, such as day time and night time.

Response: We have added the following description in the revised manuscript:

The aerosol retrievals from MISR and MODIS are only available for clear-sky conditions in the daytime. CALIPSO provides retrievals during both day and night, but only clear-sky daytime profiles are used in order to be consistent with the products from MISR and MODIS. (Line 99-102)

Line 158: I would suggest "variations of AODs with heights below 200 m"

Response: Done. Thank you! (Line 155)

Line 162-167: The different time representation errors could be noticeable for monthly average. For example, Wang and Zhao (2017, JGR) showed that the MODIS cloud time representation errors could be up to 3-4% for monthly average (10 years data) while much smaller for yearly average and much larger for daily average. However, this might not affect the findings/conclusions of this study. You may simply indicate/inform the noticeable time representation error of 3-4% (Wang and Zhao, 2017) while what they studied are cloud coverage instead of aerosols.

Response: In the revised manuscript, we have investigated the impact of spatiotemporal sampling on seasonal variations of AOD using two sensitivity scenarios, and cited this reference. The added text is shown as follows:

As described in Section 2.1, MODIS provides near-daily global coverage but MISR and CALIPSO do not. As a result, the monthly mean AOD from different sensors is calculated based on different sets of days, which might lead to uncertainties in the estimation of monthly mean AOD (Colarco et al., 2014; Wang and Zhao, 2017). To rule out the impact of spatio-temporal sampling on seasonal variation patterns, we design two sensitivity cases: a “MODIS/Terra\_match MISR” case in which the monthly mean AOD of MODIS/Terra is calculated using only the days when MISR overpasses, and a “MODIS/Aqua\_match CALIPSO” case in which the monthly mean AOD of MODIS/Aqua is calculated using only the overpassing days of CALIPSO. The results are illustrated in Fig. 2. In all three regions, the monthly mean AODs are slightly different for “MODIS/Terra” and “MODIS/Terra\_match MISR”, but the seasonal variation patterns are largely the same. The same results are found for “MODIS/Aqua” and “MODIS/Aqua\_match CALIPSO”. As such, we conclude that sampling has little effect on the AOD seasonal variation patterns reported in this study. In fact, this conclusion is compatible with the findings of Colarco et al. (2014). Colarco et al. (2014) revealed that the spatial sampling artifacts were significant for fine aggregation grid (e.g., 0.5°), but they are reduced at coarse grid scales (e.g., 10°). In this study, we use only the mean AOD over three large regions (about 20°×20°) across 10 years, therefore the sampling artifacts are expected to be even smaller. (Line 208-224)

Line 267-269: I highly agree with this analysis. By doing this, the effects of PBL and relative humidity could be minimized.

Response: Thank you!

Line 303-304: Yang et al. (2018, AR) also showed the winter vs summer patterns (high in winter, low in summer) of inter-regional transport (between pearl river delta and Hongkong) of aerosols; Garrett et al. (2010, Tellus B) also showed the strong transport in winter and weak transport in fall for aerosols from mid-latitudes to Arctic. Actually, the vertical distribution of aerosols is very interesting in the Arctic, with high values at heights around 2-7 km, mainly caused by long-range transport from other regions such as mid-latitudes.

Response: Thank you for providing these useful references. We have included them to support the point that the seasonal variations of AOD at different vertical levels are influenced by the seasonal patterns of inter-regional transport of aerosols. (Line 328-331)

Line 374, acts -> act, strengthens -> strengthen

Response: Done. Thank you! (Line 401-403)

Line 400-405, Since AOD includes the impacts of relative humidity, is it possible that relative humidity also contributes to the diurnal variation of AOD? Another possible contribution might be

the growth of fine aerosols, as indicated by Zhao et al. (2018, AAS), the growth rates of fine mode aerosols generally starts from the morning time with a growth rate of around 2-7 nm/hour (1.7-6.5 nm/hour near Beijing region in summer). They could become accumulation mode from fine mode aerosols at night time or on next day (then day time) making AOD larger.

Response: Thank you for your suggestions. We have removed the section about diurnal variations in the revised manuscript following the 2<sup>nd</sup> reviewer's comments. We hope this is acceptable for you.

Line 462-465: I agree that absorbing aerosols could play different roles to convection and clouds from non-absorbing aerosols. However, I think both absorbing aerosols and non-absorbing aerosols will reduce the direct solar radiation reaching the surface, causing surface cooler and further affect convection and clouds in similar way (of course, no absorption of solar radiation in the air), while the effects could be weaker than absorbing aerosols.

Response: Following the reviewer's comment, we have revised these sentences as follows:

Both absorbing and non-absorbing aerosols could invigorate deep convection by serving as cloud condensation nuclei and affect convection by reducing downward solar radiation and causing surface cooling (Rosenfeld et al., 2008). However, absorbing aerosols play unique roles in convection and cloud development by heating the atmosphere. This inhibits convection in most situations (Ramanathan et al., 2005; Massie et al., 2016; Zhao et al., 2018a) but may enhance convection and cloud formation above the PBL (Wang et al., 2013; Bond et al., 2013), depending on the vertical distribution of absorbing aerosols. (Line 445-452)

Reviewer 2:

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

Summary: The paper combines retrievals and observations from multiple satellites (active and passive) and ground based (in-situ and remote sensed), in order to characterize the seasonal and diurnal variations of aerosol properties in three heavily-populated regions (EUS, WEU and ECC). The aerosols are separated into lower (< 800 m) and higher levels (> 800m), monthly averages are calculated, and annual cycles plotted. Analysis and interpretation and some speculation are presented. The main conclusions are that in all three regions, column AOD and higher level AOD all peak in the summer, whereas AOD in lower levels peaks during the winter. AOD from fine-sized particles peaks in the spring/summer and is attributed to anthropogenic sources. Dust and sea-salt peaks in the winter in WEU but are nearly constant in the other two regions. There appears to be larger nighttime/daytime AOD difference in summer than winter.

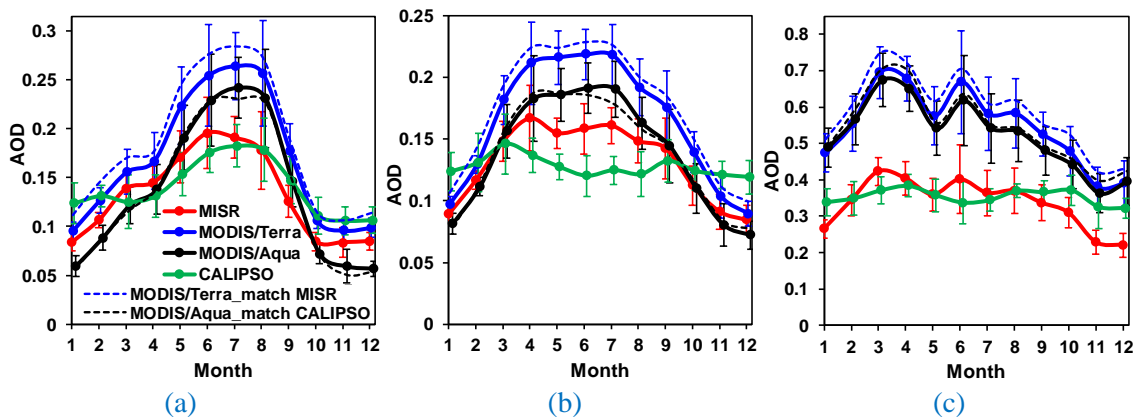
Response: We thank the reviewer for the insightful comments. We have carefully addressed these comments in revising the manuscript. Point-to-point responses are given below.

This paper is logical and easy to read. The English language usage is satisfactory. The idea of separating into low level (e.g. < 800 m; presumably a proxy for boundary layer) and higher level

(> 800 m) is unique. I wish I could believe all of the conclusions. But I don't yet. Like I explained in the "initial quality" review, I have strong concerns about data sampling. For example, Colarco et al., (2014, [10.5194/amt-7-2313-2014]) explains that "sampling matters", and that when we develop climatology from different types of orbits (and coverage), we need to deal with this problem. Because of this, I don't think that "the impact of the sampling issue is expected to be much smaller than that on the AOD retrieval in an individual month at a specific location" (Lines 162-167). If MODIS calculates monthly mean based on all 30 days and CALIPSO based on 2-4 times month (every 8 or 16 days, if lucky), then we don't expect the monthly means to match. Of course, if there are clouds, this could be MODIS making monthly means from, say 10 days, and CALIPSO making only one. One more paper to think about is Chin et al., (2014; [10.5194/acp-14-3657-2014]). Although they study multi-year data, they make points about comparing datasets with all kinds of sampling differences.

Response: We thank the reviewer for this valuable comment. To investigate the impact of data sampling on seasonal variation of AOD, we design two sensitivity cases: a "MODIS/Terra\_match MISR" case in which the monthly mean AOD of MODIS/Terra is calculated using only the days when MISR overpasses, and a "MODIS/Aqua\_match CALIPSO" case in which the monthly mean AOD of MODIS/Aqua is calculated using only the overpassing days of CALIPSO. The results are illustrated in the following figure (Fig. 2 in the revised manuscript). In all three regions, the monthly mean AODs are slightly different for "MODIS/Terra" and "MODIS/Terra\_match MISR", but the seasonal variation patterns are largely the same. The same results are found for "MODIS/Aqua" and "MODIS/Aqua\_match CALIPSO". As such, we conclude that sampling has little effect on the AOD seasonal variation patterns reported in this study. In fact, this conclusion is compatible with the findings of Colarco et al. (2014). Colarco et al. (2014) revealed that the spatial sampling artifacts were significant for fine aggregation grid (e.g., 0.5°), but they are reduced at coarse grid scales (e.g., 10°). In this study, we use only the mean AOD over three large regions (about 20°×20°) across 10 years, therefore the sampling artifacts are expected to be even smaller.

We have added the preceding discussions in the revised manuscript. (Line 208-224)



**Figure.** Monthly mean AOD observed by MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. For CALIPSO, only clear-sky daytime profiles are averaged in order to be consistent with the MISR and MODIS products. "MODIS/Terra\_match MISR" is a sensitivity case in which the monthly mean AOD of MODIS/Terra is calculated using only the days when MISR overpasses, and "MODIS/Aqua\_match

CALIPSO” is a case in which the monthly mean AOD of MODIS/Aqua is calculated using only the overpassing days of CALIPSO. The error bars denote the standard deviation of the monthly mean AOD values obtained over all years. Note the different scales on the y-axes of the plots.

Of course, the low bias (Fig 2, lines 216-218) between CALIPSO -derived AOD and the other satellites (MODIS, MISR), can be because of assumed lidar ratios (Ma et al., AMT; [10.5194/amt-6-2391-2013]), or undetected aerosol layers (Kim et al., JGR, [10.1002/2016JD025797]). I guess that the Thorson (2017) reference already listed could be a reason as well.

Response: Thank you! We have added these explanations to the revised manuscript, citing the two references. (Line 315-318)

Why the non-confident statements (“probably”) in lines 227-231? I think you should be able to find references. What are the sources of these SOA? What about long-range transport in the summer?

Response: To address the reviewer’s comments, we have revised these sentences as follows:

Considering the high accuracy of AERONET, we conclude that AOD peaks in summer/spring and dips in winter. An important reason for the higher AOD in summer is that the stronger radiation and higher temperature accelerate the formation of secondary aerosols (Timonen et al., 2014), including sulfate, nitrate, ammonium, and secondary organic aerosol (SOA). SOA is produced by photo-oxidation of volatile organic compounds (VOCs) and intermediate volatility organic compounds (IVOCs), as well as the chemical aging of primary organic aerosol (Zhao et al., 2016). Another reason is that more abundant water vapor in summer favors the hygroscopic growth of aerosols (Liu et al., 2012; Zheng et al., 2017). The different patterns of long range transport as a function of season is also partly responsible for the seasonable variation of AOD (Tian et al., 2017; Yang et al., 2018; Garrett et al., 2010). (Line 232-242)

Considering lines 235-236, I again ask about sampling? Are the monthly means for AERONET and satellites computed based on the same days? Or is mean AERONET = mean (of AERONET data) and mean satellite = mean (of SATELLITE data)? We know from validation exercises that when actually collocated in space and time (both AERONET and satellite are free of clouds) that they match overall well (yes, sometimes small biases, e.g. Remer et al., 2005). However, I do not expect matches if using different samples (days). Note that the Remer et al., (2005) study has been updated for MODIS (Collection 5 and Collection 6), and there are also updates for MISR evaluations. The “instrument calibration issue” (lines 243-244) would not cause such a large bias.

Response: For each satellite-borne sensor, only those days for which the satellite overpasses an AERONET site were used in the comparisons. In other words, the monthly means for AERONET and satellites were indeed computed based on the same days. In addition, to match coincident measurements, the AERONET AOD retrievals for each site were averaged within a 2 h window centered on the satellite overpass times (about 10:30 for MISR and MODIS/Terra, and 13:30 for MODIS/Aqua and CALIPSO, depending on site location), and compared with the satellite AOD retrievals in a  $1^\circ \times 1^\circ$  grid box that contains the corresponding AERONET site. (Line 172-177)

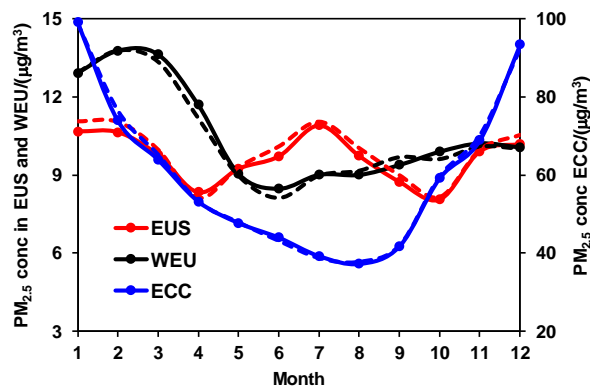


Thank you for pointing out that the Remer et al. (2005) study is outdated. For MISR, however, the Kahn et al. (2010) study is applicable to the product used in this paper. We have revised the descriptions about the discrepancies among MODIS, MISR and AERONET as follows:

While relative patterns of AOD seasonal variations from observations of MISR, MODIS/Terra, and MODIS/Aqua are similar to each other and to those of AERONET, the magnitude of AOD observed by these sensors shows considerable discrepancies. In all three regions, the AOD retrieved from MODIS is larger than that from MISR, consistent with the results of previous studies (de Meij et al., 2012; Zhao et al., 2017; Chin et al., 2014; Kang et al., 2016; Qi et al., 2013). This is most likely due to differences in observing strategy, retrieval algorithms, and spatio-temporal sampling (Kahn et al., 2009). The MISR-retrieved AOD agrees well with the AERONET observations in EUS and WEU regions. In the ECC region, however, MISR underestimates the AERONET AOD, probably because there is less signal from the surface at higher AOD, which creates ambiguity that can result in the algorithm assigning too much of the top-of-atmosphere radiance to the surface (i.e., a higher surface albedo), thereby underestimating the AOD (Kahn et al., 2010). The MODIS/Terra and MODIS/Aqua overestimate the AERONET AOD to some extent in all three regions. The overestimation was also reported in two previous studies (de Meij et al., 2012; Ruiz-Arias et al., 2013) using the level 3 MODIS products (Collection 5 or 5.1). We show a relatively larger overestimation than that reported by de Meij et al. (2012) and Ruiz-Arias et al. (2013), partly because we used the AERONET AOD averaged within a 2 h window centered on the satellite overpass times while the two previous studies used the daily/monthly mean AERONET AOD in the comparisons. The daily mean AOD observed by AERONET is about 10% larger than the value during the satellite overpass times (Li et al., 2013). The reasons for the overestimation are yet to be thoroughly elucidated in future studies. (Line 243-262)

I think it is a good idea that you are comparing low-level CALIPSO to ground level PM<sub>2.5</sub> (lines 267-269) but I wonder about the temporal sampling. Also, PM<sub>2.5</sub> is usually a “dried” aerosol measurement whereas CALIPSO is ambient RH.

Response: In the original manuscript, the monthly mean PM<sub>2.5</sub> concentrations were calculated using observations in all days. Here we redo the calculation using only the days when CALIPSO overpasses an observational site (dashed lines in the following figure, shown below), and compare with the original estimates (solid lines). The results show that the temporal sampling has minor effects on the monthly mean PM<sub>2.5</sub> concentrations. In the revised manuscript, we used the updated calculation method (dashed lines) in order to match the CALIPSO observations.



**Figure.** Monthly mean surface PM<sub>2.5</sub> concentrations during 2007-2016 in three target regions. The solid lines represent monthly mean PM<sub>2.5</sub> concentrations calculated using observations in all days, while the dashed lines are calculated using only the days when CALIPSO overpasses an observational site. The numbers of observational sites included in averaging are 225, 52, and 496, in the EUS, WEU, and ECC regions. Note the different scales on the y-axes for EUS/WEU and ECC.

We agree with the reviewer that the low-level AOD observed by CALIPSO is affected by ambient RH. Nevertheless, previous studies have reported fairly good correlations between extinction coefficient/low-level AOD and PM<sub>2.5</sub> concentrations (Cheng et al., 2013; Zheng et al., 2017). In addition, we intend to do a qualitative and not quantitative comparison. For these reasons, it appears reasonable to compare the seasonal variation patterns of low-level AOD and PM<sub>2.5</sub> concentrations. We have included the discussions in the revised manuscript (Line 281-284).

I don't understand the arguments in lines 279-282, in that since CALIPSO can't detect thin aerosols, that the fraction of upper-level aerosols is smaller than at the surface, and that results in the CALIPSO AOD as being weighted toward lower heights. According to the Kim et al., paper (listed above), CALIPSO is likely to miss stuff close to the ground. Anyway, the point is I don't think you can say that CALIPSO is missing stuff, and yet it "provides valuable information with respect to intra-annual variations at specific height ranges" (line 289-290).

Response: Indeed, the aerosols with heights below 200 m AGL are frequently undetected because of surface contamination (Kim et al., 2017; NASA CALIPSO team, 2011), but the overall fraction of aerosols detected in the upper levels (> 800 m AGL) is still much smaller than that in the lower levels (< 800 m AGL) because the upper-level aerosols tend to be optically thin. This is evident from Fig. 10 of Kim et al. (2017) and Fig. 1 of Thorsen et al. (2017). Therefore, the CALIPSO-observed AOD seasonal variations are significantly weighted toward lower heights.

The detection sensitivities in the upper and lower levels differ significantly because the extinction coefficient decreases by about 2 orders of magnitude with an increase of height (Kim et al., 2017; Thorsen et al., 2017). Within a specific height range, however, the optical thickness of aerosols and hence the detection fraction has a smaller variability. This is supported by the fact that the seasonal mean AOD within a specific height range differs by at most 3 times as a function of season (Fig. 4 in the main text). For these reasons, we argue that CALIPSO could provide valuable information with respect to seasonal variations of aerosols within a specific height range.

We have added the preceding discussions in the revised manuscript. (Line 295-304, 309-315)

Thank you for adding many references in lines 292-304 to discuss why AOD seasonal differences should be different in lower versus higher altitudes. I don't know if I agree that "seasonal variations of AOD at different levels are influenced by variations of RH which affects hygroscopic growth" (Lines 301-303). Of course, RH influences AOD, but it is total column water vapor and not necessarily RH that changes drastically from season to season.

Response: Following the reviewer's comment, we have changed "RH" to "water vapor amount".



I agree that comparing MISR-derived aerosol “types (size/ absorption)” and CALIPSO derived aerosol “types” (sources) is ultimately useful. (lines 307-308). However, they are clearly different beasts, and I am getting lost reading this section (Section 3.3). Each paragraph has multiple sentences that are “A implies B, whereas (while / in contrast) sometimes C implies D”. It’s hard to follow. I suggest a table, or schematic cartoons, or bullets.

Response: Following the reviewer’s comment, we have added a table summarizing the seasonal variations of different aerosol types in the three regions (shown below). We have also carefully revised the text of this section improve the logic and readability. (Line 333-408)

**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

I notice that regarding the aerosol typing as seen by CALIPSO, all of the regions (over land), have non-trivial amount of “clean-marine” aerosol (Fig 7). Is this transported marine aerosol to the entirety of the regional box, or should the marine aerosol be expected to be more dominant but confined only to the coastal areas of a region?

Response: All three regions used in the study cover some ocean areas (see Fig. 1 in the manuscript). The marine aerosols are predominantly located over the ocean and in coastal areas, and are much fewer over land.

The section on daytime/nighttime variability is nice, but I think it is beside the point of the rest of the paper. Why would smoke AOD accumulate at night? Higher RH at night might make bigger aerosols, but if anything, fire activity is reduced at night. You might check the PM<sub>2.5</sub> measurements here. I suggest leaving this section out, and thinking about the questions related to the other sections. “Intra-annual”, “vertical”, and “particle types” is enough for one paper!

Response: Following the reviewer’s suggestion, we have left out this section in the revised manuscript.

In terms of figures. I can see why the authors do this (different magnitudes of AOD and or PM<sub>2.5</sub> at different sites), but the varying y-axes within figure captions, and from figure-to-figure are distracting. But thank you for pointing out in the caption!

Response: We tried to unify the scales of the y-axes but failed because the magnitude differs greatly according to figures. Thank you for your understanding.

What is “upper air”? I see it a few places, and assume you mean > 800 m AGL? (e.g. Line 300).

Response: Yes, it means > 800 m AGL. We have explained this in the revised manuscript. (Line 299)

The abstract suggests (lines 37-38) that results can “help to improve the current estimates of climatic and health impacts of aerosols”. Well maybe, but I would drop this from the abstract since there is no discussion in the paper.

Response: Following the reviewer’s comment, we have removed this sentence from the abstract.

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

33 in summer over the EUS and WEU regions, whereas they follow a bimodal distribution with peaks  
34 in August and March (due to crop residue burning) over the ECC region.

35

## 36 **1 Introduction**

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

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

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

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

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

83 In this study, we investigate the seasonal variations of aerosol column loading, vertical  
84 distribution, and particle types using multiple satellite and ground-based observational datasets  
85 covering the period from 2007 to 2016. The purpose is to assess the consistency among various  
86 datasets and provide a comprehensive characterization of aerosol properties in polluted regions to  
87 facilitate future studies of aerosol climate effects and local air quality issues. The data are from  
88 MISR, MODIS, CALIPSO, Aerosol Robotic Network (AERONET), and surface PM<sub>2.5</sub> monitors.  
89 Consistent with our previous study (Zhao et al., 2017), we selected three populous regions which  
90 have experienced substantial anthropogenic pollution (Wang et al., 2017; Wang et al., 2014) and  
91 have received considerable attention in other climate studies: the Eastern United States (EUS; 29°-  
92 45° N, 70°-98° W), Western Europe (WEU; 37°-59° N, 10° W-17° E), and Eastern and Central  
93 China (ECC; 21°-41° N, 102°-122° E). The geographical boundaries of these regions are shown in  
94 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  $0.5^\circ \times 0.5^\circ$ . The  
109 variables used in the analysis are total AOD at 555 nm as well as AODs for six aerosol components,  
110 namely small ( $< 0.7 \mu\text{m}$  diameter), medium (0.7-1.4  $\mu\text{m}$  diameter), large ( $> 1.4 \mu\text{m}$  diameter),  
111 spherical, non-spherical, and absorbing. Based on comparison with ground-based AERONET  
112 measurements, the errors in MISR AOD data are on the order of  $\pm 0.05$  or  $\pm(0.20 \times \text{AOD})$ ,  
113 whichever is larger (Kahn et al., 2005; Kahn et al., 2010). In addition, retrieval of MISR particle  
114 property information from individual retrievals is considered to be reliable when  $\text{AOD} > 0.15$ , and  
115 has diminished sensitivity at smaller AOD (Kahn and Gaitley, 2015; Kahn et al., 2010). In this  
116 study we use only monthly mean values, for which the uncertainties are expected to be smaller  
117 than those for individual retrievals. Note that we did not do a relative humidity (RH) correction to  
118 AOD retrievals from MISR as well as other sensors. The seasonal variations of AOD represent a  
119 combined effect of variations in aerosol abundance, vertical distribution, chemical constituents,  
120 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 AOD retrievals to be about  $\pm(0.05 + 0.15 \times \text{AOD})$  over land and  $\pm(0.03 + 0.05 \times \text{AOD})$  over  
127 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<sub>2.5</sub> 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.

225 In view of the substantial differences between CALIPSO and the other three sensors, we  
226 compare satellite retrieved AOD seasonal variations with point-based ground measurements from  
227 AERONET (Fig. 3). As in other studies, AERONET data are treated as “ground truth” for column  
228 AOD due to its smaller uncertainty compared with satellite data (Kahn et al., 2010; Levy et al.,  
229 2010; Schuster et al., 2012; Fan et al., 2018). Figure 3 shows that, in all three regions, the AOD  
230 seasonal variations measured by AERONET are similar to those retrieved by MISR, MODIS/Terra,  
231 and MODIS/Aqua, but are quite different from CALIPSO data. Reasons for the different seasonal  
232 patterns between CALIPSO and other sensors will be discussed in Section 3.2. Considering the  
233 high accuracy of AERONET, we conclude that AOD peaks in summer/spring and dips in winter.  
234 An important reason for the higher AOD in summer is that the stronger radiation and higher  
235 temperature accelerate the formation of secondary aerosols (Timonen et al., 2014), including  
236 sulfate, nitrate, ammonium, and secondary organic aerosol (SOA). SOA is produced by photo-  
237 oxidation of volatile organic compounds (VOCs) and intermediate volatility organic compounds  
238 (IVOCs), as well as the chemical aging of primary organic aerosol (Zhao et al., 2016). Another

239 reason is that more abundant water vapor in summer favors the hygroscopic growth of aerosols  
240 (Liu et al., 2012; Zheng et al., 2017). The different patterns of long range transport as a function  
241 of season is also partly responsible for the seasonable variation of AOD (Tian et al., 2017; Yang  
242 et al., 2018; Garrett et al., 2010).

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

261 AERONET is about 10% larger than the value during the satellite overpass times (Li et al., 2013).  
262 The reasons for the overestimation are yet to be thoroughly elucidated in future studies.

### 263 3.2 Seasonal variations of aerosol loadings as a function of height

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

275 To provide an independent evaluation of the CALIPSO-observed AOD variations at lower  
276 heights, we examine the seasonal variations of near-surface PM<sub>2.5</sub> concentrations at hundreds of  
277 surface monitor locations within the three target regions (Figure 5). The aerosol extinction  
278 coefficient, and hence AOD at lower heights is affected by not only the particle mass  
279 concentrations, but also aerosol type (absorbing vs. nonabsorbing aerosols, coarse-mode vs. fine-  
280 mode aerosols) and meteorological parameters such as RH, wind speed and direction, and  
281 planetary boundary layer height (Zheng et al., 2017). Nevertheless, previous studies have reported  
282 fairly good correlations between extinction coefficient/low-level AOD and PM<sub>2.5</sub> concentrations  
283 (Cheng et al., 2013; Zheng et al., 2017). For this reason, it is reasonable to qualitatively compare



284 the seasonal variation patterns of near-surface PM<sub>2.5</sub> concentrations and low-level AOD. We  
285 calculate monthly mean PM<sub>2.5</sub> concentrations using only the days when CALIPSO overpasses an  
286 observational site to enable a better comparison. Figure 5 shows that, over the ECC and WEU  
287 regions, surface PM<sub>2.5</sub> concentrations are largest in winter and smallest in summer. In the EUS  
288 region, the maximum PM<sub>2.5</sub> concentration occurs in summer and a second maximum occurs in  
289 winter. These trends are generally consistent with the seasonal variations of AOD at low heights,  
290 implying that CALIPSO data can generally capture the seasonal changes in low-level aerosol  
291 abundance.

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

307 For the EUS regions, due to the bimodal AOD distribution at lower heights, the summer peak in  
308 column AOD variations remain but the difference between peak and valley is smaller than implied  
309 by the observations of MISR/MODIS/AERONET. Although the integrated CALIPSO column  
310 AOD does not agree well with AERONET, it does provide valuable information with respect to  
311 seasonal variations of aerosols within a specific height range, since the optical thickness and hence  
312 the detection fraction has a smaller variability at a given height. This is supported by the fact that  
313 the seasonal mean AOD within a specific height range differs by at most 3 times as a function of  
314 season (Fig. 4), while it decreases by about 2 orders of magnitude with an increase of height (Kim  
315 et al., 2017; Thorsen et al., 2017). Besides the seasonal variations, the difference in the magnitude  
316 of AOD between CALIPSO and other sensors are also largely explained by the undetected aerosol  
317 layers by CALIPSO (Kim et al., 2017; Thorsen et al., 2017) as well as the assumed lidar ratios in  
318 CALIPSO retrievals (Ma et al., 2013).

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

330 et al., 2017) as well as the seasonal patterns of inter-regional transport of aerosols (Tian et al.,  
331 2017; Yang et al., 2018; Garrett et al., 2010).

### 332 3.3 Seasonal variations of aerosol types

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

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

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

363 With regard to dust and clean marine (sea-spray) aerosols, the AOD in the EUS region does  
364 not show an obvious seasonal pattern. In the WEU region, AOD of dust aerosols peaks in summer,  
365 consistent with previous surface-based observational studies which show that dust events in  
366 Europe predominantly occur during summer due to transport from the Sahara region (Stafoggia et  
367 al., 2016). The AOD of dust is primarily located above 800 m, supporting the conclusion that dust  
368 aerosols in WEU mainly originate from long range transport. Since the dust AOD is subject to a  
369 large inter-annual variability (denoted by the large error bars in Fig. 7), we use the Student's t-test  
370 to demonstrate the statistical significance of the seasonal variations. The dust AOD in summer is  
371 statistically larger than that in any other season at the 0.05 level, indicating the robustness of the  
372 peak in summer. Contrary to dust, the AOD of sea-spray aerosols in WEU is much higher in winter  
373 than in summer, probably because winter is the relative windy season with large low pressure  
374 systems over the Atlantic Ocean and the North Sea (Manders et al., 2009). The offset of the  
375 opposite variation trends in dust and sea-spray aerosols partly accounts for the rather uniform

376 distributions of large-size AOD in WEU (see Fig. 6). Over the ECC region, sea-spray aerosols  
377 make a negligible contribution to total AOD. The dust AOD is much larger in spring than in any  
378 other season (significant at the 0.05 level), which is tied to the outburst of springtime Gobi desert  
379 dust storms (China Meteorological Administration, 2012). The high dust AOD explains the peak  
380 in large-size AOD in spring over the ECC region (see Fig. 6).

381 Smoke aerosols are predominantly located above 800 m in all regions. Over the EUS and  
382 WEU regions, smoke aerosols present a unimodal distribution with maximum occurring in summer.  
383 The differences between smoke AOD in summer and the other three seasons are all statistically  
384 significant at the 0.05 level, except for the difference between summer and spring over the WEU  
385 region, which is statistically significant at the 0.10 level. In the ECC region, the smoke AOD  
386 follows a bimodal distribution with peaks occurring in March and August and valleys occurring in  
387 May and December. The differences between either of the peak months and either of the valley  
388 months are statistically significant at the 0.05 level. MISR's independent retrieval of absorbing  
389 AOD (Fig. 6) presents a highly similar seasonal pattern (statistically significant at the 0.05 level)  
390 as the CALIPSO smoke AOD. In fact, smoke and absorbing aerosols are closely correlated with  
391 each other, since smoke consists of a much larger fraction of absorbing aerosols (Dubovik et al.,  
392 2002), such as BC and light-absorbing organic aerosol (Kirchstetter and Thatcher, 2012), as  
393 compared to other aerosol types. Besides, the MISR absorbing AOD and CALIPSO smoke AOD  
394 are also consistent in the order of magnitude. The variability of MISR absorbing AOD (shown in  
395 the right Y-axis of Fig. 6) is about 0.002-0.005, while the variability of smoke AOD from  
396 CALIPSO is about 0.01-0.03. The smoke AOD includes the contributions of both the absorbing  
397 and scattering portions. The MISR absorbing AOD, which is calculated using  $\text{total AOD} \times (1 -$   
398  $\text{single scattering albedo})$ , represents only the absorbing portion but includes contributions from

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

#### 409 **4 Conclusions and implications**

410 This study investigated the seasonal variations of aerosol column loading, vertical  
411 distribution, and particle types using multiple satellite and ground-based observational datasets  
412 during 2007-2016 over EUS, WEU, and ECC regions. Retrievals from MISR and MODIS reveal  
413 that column AOD in all three regions peaks in spring/summer and reaches its low in winter, which  
414 is consistent with observations from AERONET. This seasonal pattern is probably explained by  
415 accelerated formation of secondary aerosols in spring/summer due to stronger insolation and  
416 higher temperature. In contrast, CALIPSO shows a much weaker seasonal variability in column  
417 AOD, probably because CALIPSO-retrieved AOD is weighted toward lower heights since some  
418 thin aerosol layers in high levels are undetected due to insufficient detection sensitivity. Despite  
419 the discrepancy in integrated column AOD, CALIPSO does provide valuable information with  
420 respect to intra-annual variations of AOD as a function of height. Over the WEU and ECC regions,  
421 AODs of the vertical layers below 800 m generally peak in winter, while those above 800 m mostly



422 peak in summer. For the EUS region, the maximum AOD above 800 m also occurs in summer;  
423 however, AOD below 800 m shows two peaks, one in summer and the other in winter. The seasonal  
424 variations of AOD at low heights are consistent with seasonal patterns of measured surface PM<sub>2.5</sub>  
425 concentrations.

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

438         The combination of multiple satellite and ground-based observations facilitate a systematic  
439 and deeper understanding of the seasonal variations of aerosols, particularly their vertical and type  
440 distribution. Comparison of multiple measurement and retrieval methodologies enables reducing  
441 the uncertainties in the estimation of aerosol direct effects by providing improved information  
442 about aerosol vertical and type distributions, which significantly affect the aerosol-induced  
443 scattering and absorption of radiation. More importantly, the intra-annual variations of vertical  
444 distributions and types of aerosols are important for understanding their impact on atmospheric

445 dynamics, cloud fields, and precipitation production. For example, both absorbing and non-  
446 absorbing aerosols could invigorate deep convection by serving as cloud condensation nuclei and  
447 affect convection by reducing downward solar radiation and causing surface cooling (Rosenfeld  
448 et al., 2008). However, absorbing aerosols play unique roles in convection and cloud development  
449 by heating the atmosphere. This inhibits convection in most situations (Ramanathan et al., 2005;  
450 Massie et al., 2016; Zhao et al., 2018a) but may enhance convection and cloud formation above  
451 the PBL (Wang et al., 2013; Bond et al., 2013), depending on the vertical distribution of absorbing  
452 aerosols. Finally, the data and variation patterns presented in this study can be used to evaluate  
453 and improve model simulations, with the ultimate goal of improving model assessment of the  
454 climatic and health effects of aerosols.

455

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461

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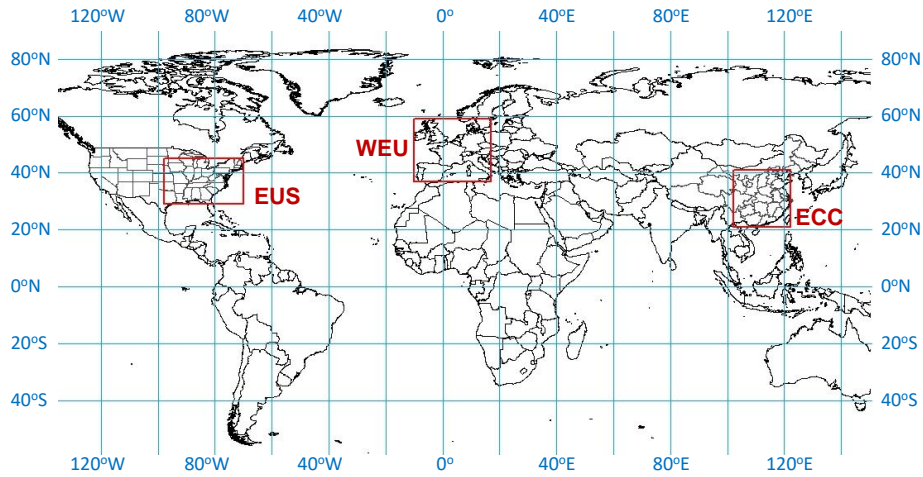
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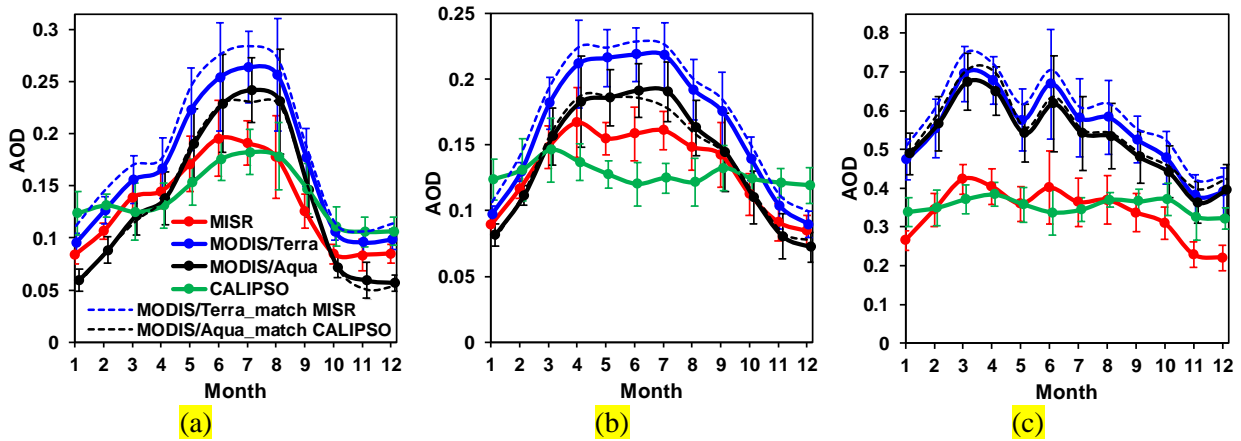


721 **Figures**



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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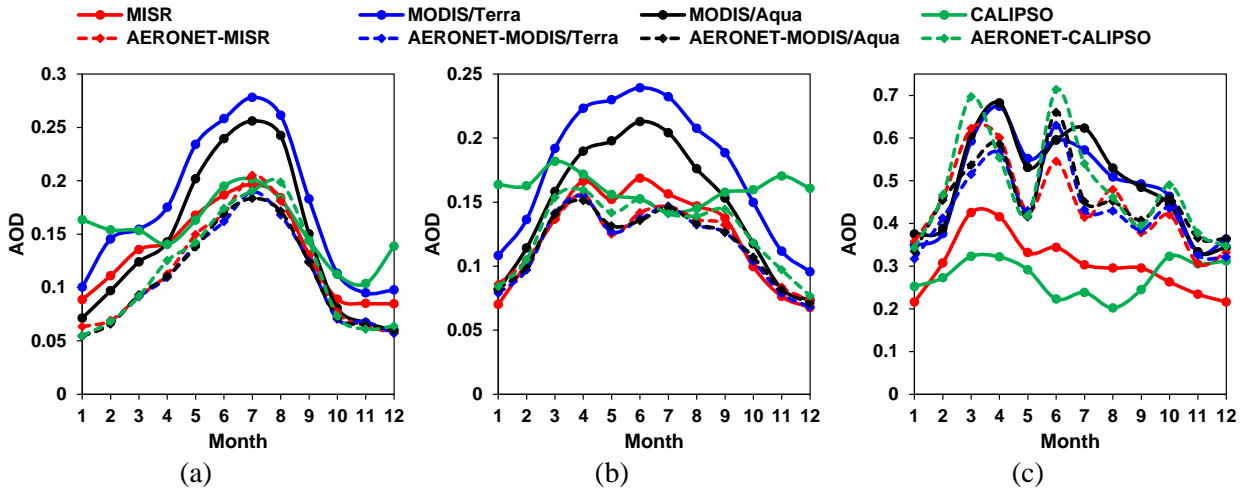


726 **Figure 2.** Monthly mean AOD observed by MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO  
727 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. For CALIPSO, only clear-sky daytime  
728 profiles are averaged in order to be consistent with the MISR and MODIS products.  
729 “MODIS/Terra\_match MISR” 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

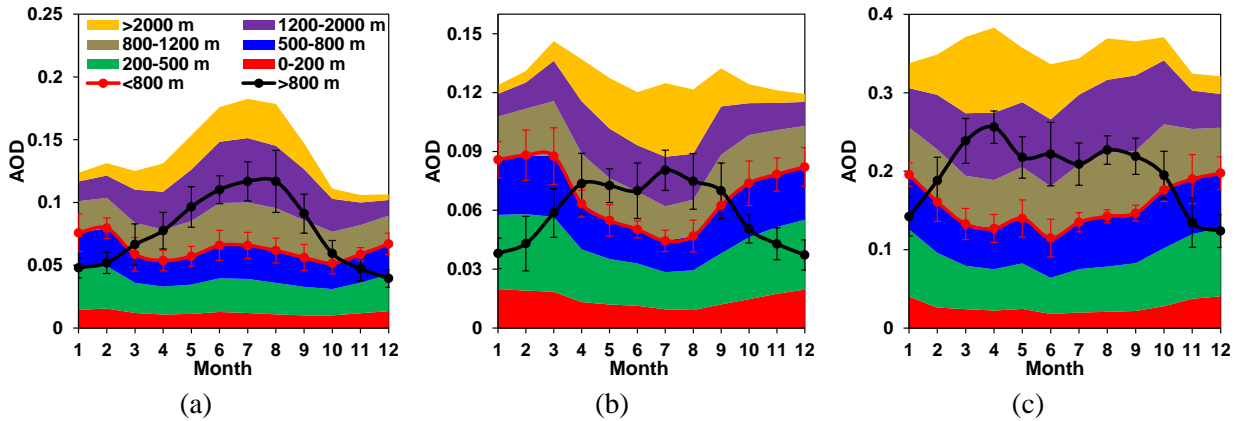


732 overpassing days of CALIPSO. The error bars denote the standard deviation of the monthly mean  
 733 AOD values obtained over all years. Note the different scales on the y-axes of the plots.

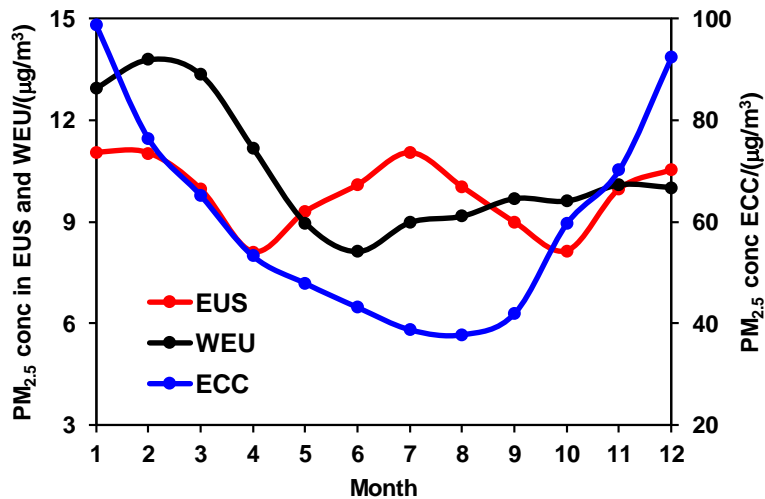
734



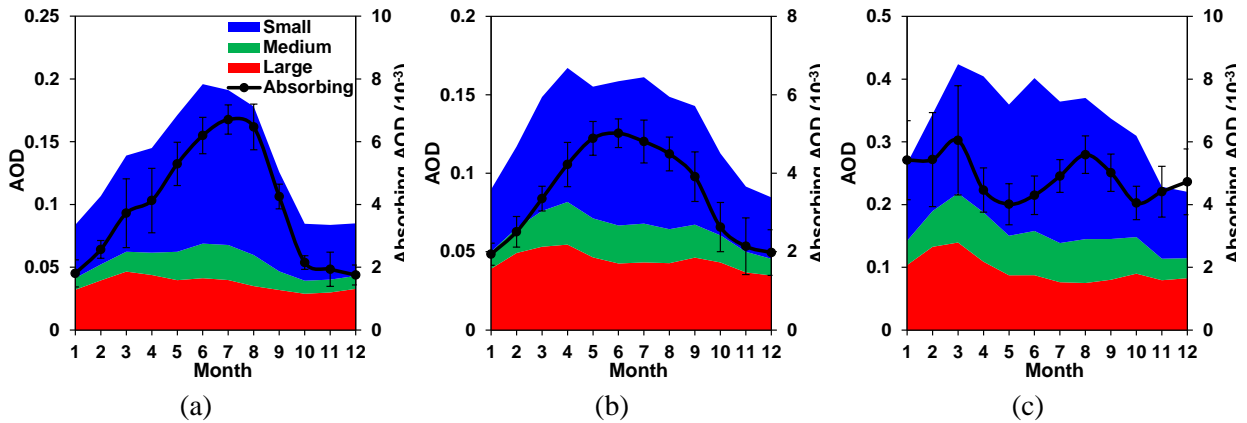
735 **Figure 3.** Monthly mean AOD observed by satellites and AERONET averaged across the  
 736 AERONET sites during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. The observations from  
 737 MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO are averaged over  $1^{\circ} \times 1^{\circ}$  grid boxes containing  
 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.



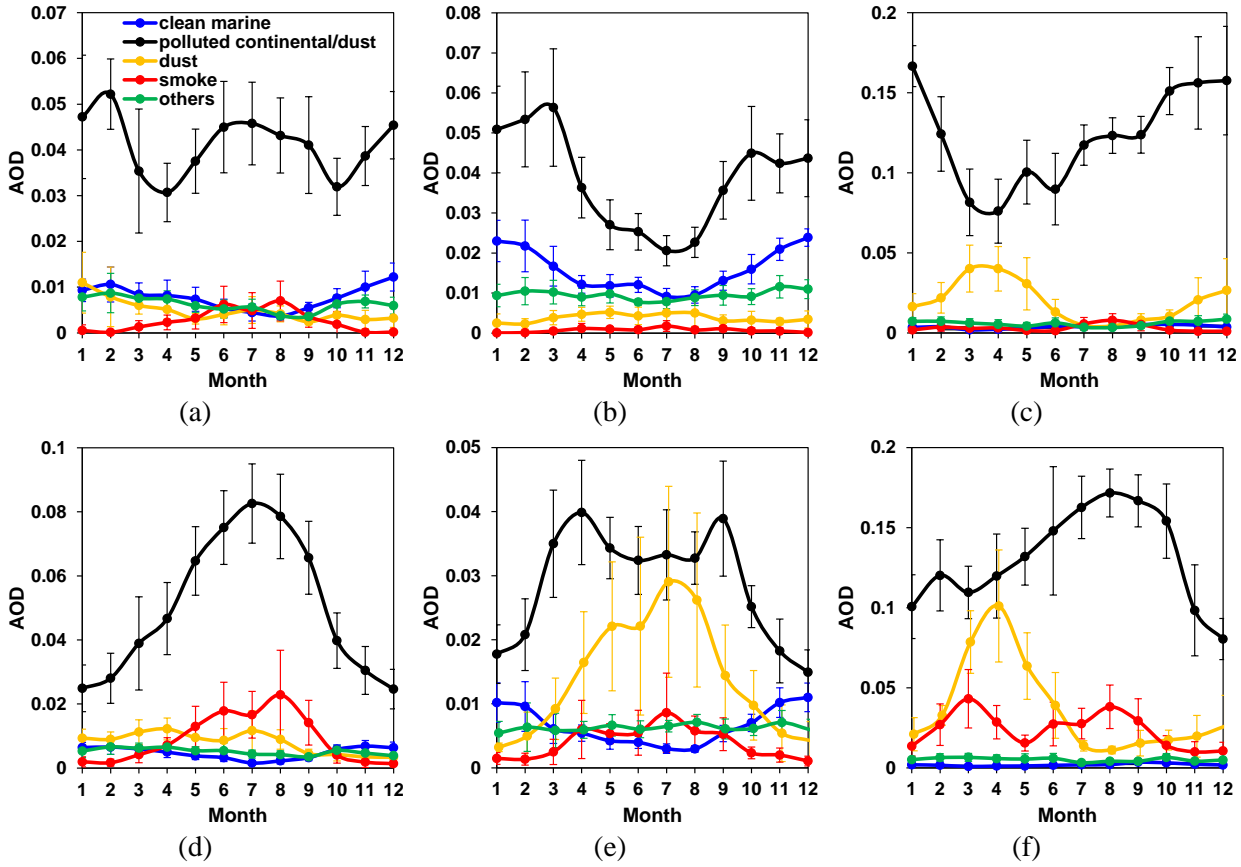
747 **Figure 4.** Monthly mean AOD as a function of height above ground level observed by CALIPSO  
 748 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. Only clear-sky daytime profiles are averaged  
 749 in order to be consistent with the products of MISR and MODIS. The range of AOD within a  
 750 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>
<b>Total column AOD</b>	Peak in summer	Peak in summer/late spring	Peak in summer/spring
<b>AOD &gt; 800 m AGL</b>	Peak in summer	Peak in summer/late spring	Peak in summer/spring
<b>AOD &lt; 800 m AGL</b>	Two peaks in winter and summer	Peak in winter	Peak in winter
<b>Small-size</b>	Peak in summer	Peak in summer/late spring	Peak in summer/spring
<b>Medium-size</b>	Peak in summer	Peak in summer/late spring	Peak in summer/spring
<b>Large-size</b>	Rather uniform	Rather uniform	Peak in spring
<b>Absorbing</b>	Peak in summer	Peak in summer/late spring	Two peaks in Mar and Aug
<b>Polluted continental/dust</b>	Similar to height-specific total AOD	Similar to height-specific total AOD	Similar to height-specific total AOD
<b>Dust</b>	No obvious seasonal pattern	Peak in summer	Peak in spring
<b>Clean marine</b>	No obvious seasonal pattern	Peak in winter	Negligible amount
<b>Smoke</b>	Peak in summer	Peak in summer/late spring	Two peaks in Mar and Aug

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