

Satellite-based estimate of the variability of warm cloud properties associated with aerosol and meteorological conditions

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Abstract. Aerosol-cloud interaction is examined using ten years of data from the MODIS/Terra (morning orbit) and MODIS/Aqua (afternoon orbit) satellites. Aerosol optical depth (AOD) and cloud properties retrieved from both sensors are used to explore in a statistical sense the morning-to-afternoon variation of cloud properties in conditions with low and high AOD, over both land and ocean. The results show that the interaction between aerosol particles and clouds is more complex and of greater uncertainty over land than over ocean. The variation in $d(\text{Cloud}_X)$, defined as the mean change in cloud property Cloud_X between the morning and afternoon overpasses in high AOD conditions minus that in low AOD conditions, is different over land and ocean. This applies to cloud droplet effective radius (CDR), cloud fraction (CF) and cloud top pressure (CTP), but not to cloud optical thickness (COT) and cloud liquid water path (CWP). Both COT and CWP increase over land and ocean after the timestep, irrespective of the AOD. However, the initial AOD conditions can affect the amplitude of variation of COT and CWP. The effects of initial cloud fraction and meteorological conditions on the change in CF under low and high AOD conditions after the 3 hours timestep over land are also explored. Two cases are considered: (1) when the cloud cover increases; (2) when the cloud cover decreases. For both cases, we find that almost all values of $d(\text{CF})$ are positive, indicating that the variations of CF are larger in high AOD than that in low AOD after the 3 hours timestep. The results also show that large cloud fraction occurs when scenes experience large AOD and stronger upward motion of air parcels. Furthermore, scenes with large cloud fraction experience large AOD and larger RH when RH is larger than 20%. We also find that smaller cloud fraction occurs when scenes experience larger AOD and larger initial cloud cover. Overall, the analysis of the diurnal variation of cloud properties provides a better understanding of aerosol-cloud interaction over land and ocean.

Key words: MODIS, cloud development, aerosol-cloud interaction, urban clusters, ocean

1 Introduction

Clouds and cloud systems are crucial elements in the energy cycle of our planet (Hartmann et al., 1992; Webb et al., 2006). Clouds affect the global energy budget by reflecting incoming solar radiation, and thus cool the Earth surface, and by absorption and re-emitting outgoing terrestrial radiation which contributes to

1 warming of the surface. In addition to the radiative effects, clouds also influence the hydrological cycle of
2 the Earth through precipitation (Stephens et al., 2002). Due to interactions with aerosols, the climatic effects
3 of clouds are further complicated (Rosenfeld, 2000; Twomey, 1974; Twomey, 1977). Aerosols can serve as
4 cloud condensation nuclei (CCN), depending on their hygroscopic properties, and when activated they can
5 change the cloud microphysical properties. The increase of CCN, while the liquid water path remains
6 constant, usually results in more numerous cloud droplets with smaller cloud droplet radius (CDR) due to
7 the competition for the same amount of water vapour. Thus, cloud albedo increases and the smaller cloud
8 droplet effective radius in most cases results in the suppression of precipitation, which in turn results in a
9 longer cloud lifetime, and maintaining a larger liquid water path (Albrecht, 1989; Feingold et al., 2001).
10 Therefore, it is important to understand the interaction between aerosols and clouds and the effect of
11 different processes on cloud development.

12 Numerous studies have shown that aerosol particles can affect cloud properties on regional and global scales
13 (Krüger and Graßl, 2002; Menon et al., 2008; Rosenfeld et al., 2014; Sporre et al., 2014; Saponaro et al.,
14 2017). Satellite measurements suggest that the cloud droplet effective radius (CDR) decreases with
15 increasing aerosol optical depth (AOD, which is used in this paper as a proxy for aerosol concentration),
16 which is consistent with Twomey's theory (Kaufman et al., 2005; Matheson et al., 2005; Meskhidze and
17 Nenes, 2010). However, other observational and model studies reported that CDR tends to increase with
18 aerosol loading in some study areas, especially over land (Feingold et al., 2001; Yuan et al., 2008; Grandey
19 and Stier, 2010; Liu et al., 2017). A different behaviour of cloud cover as a function of AOD for different
20 aerosol loadings (low or high) has been found by Kaufman and Koren (2006) and Koren et al. (2008).
21 However, the observed correlations between aerosol and cloud cannot be simply attributed to the effects of
22 aerosols on clouds alone since other factors such as variations in meteorological conditions could play a role
23 (Loeb and Schuster, 2008; Reutter et al., 2009; Koren et al., 2010; Su et al., 2010; Stathopoulos et al., 2017).
24 "Snapshot" studies, where the aerosol and cloud properties are retrieved at the same time, have the
25 advantage that they represent the total time-integrated effect of aerosols on cloud properties (Meskhidze et
26 al., 2009; Gryspeerdt et al., 2014). However, the use of "snapshot" correlations is limited to a single
27 overpass time and limits the ability to distinguish aerosol-cloud interactions from meteorological
28 covariation or retrieval errors (Gryspeerdt et al., 2014). Therefore, the history of meteorological forcing is
29 an important determinant of cloud state. Matsui et al. (2006) investigated the properties of low clouds
30 derived from semiglobal observations by the Tropical Rainfall Measurement Mission (TRMM) and
31 explored the correlations of these cloud properties with aerosols (as indicated by the aerosol index or AI)
32 and with lower-tropospheric stability (LTS) on a diurnal scale. They found that aerosols affect the CDR
33 stronger for low LTS than for high LTS. Mauger and Norris (2007) used MODIS/Terra data to examine the
34 evolution of marine boundary layer clouds over several days but they may have missed important effects
35 occurring on a sub-daily timescale. Meskhidze et al. (2009) investigated the evolution of cloud properties
36 between the MODIS/Terra and MODIS/Aqua overpasses as a function of MODIS/Terra AOD and found an
37 apparent increase in the breakup rate of stratocumulus clouds in high AOD environments. However, they
38 did not explain meteorological covariation that may generate spurious correlations.

39 Considering the complex aerosol composition and increasing aerosol trend during the last decades over
40 eastern China (Guo et al., 2011), a systematic assessment of the effect of aerosols on the properties of warm
41 clouds is needed, over both land and ocean. In this paper, aerosol-cloud interaction is examined using

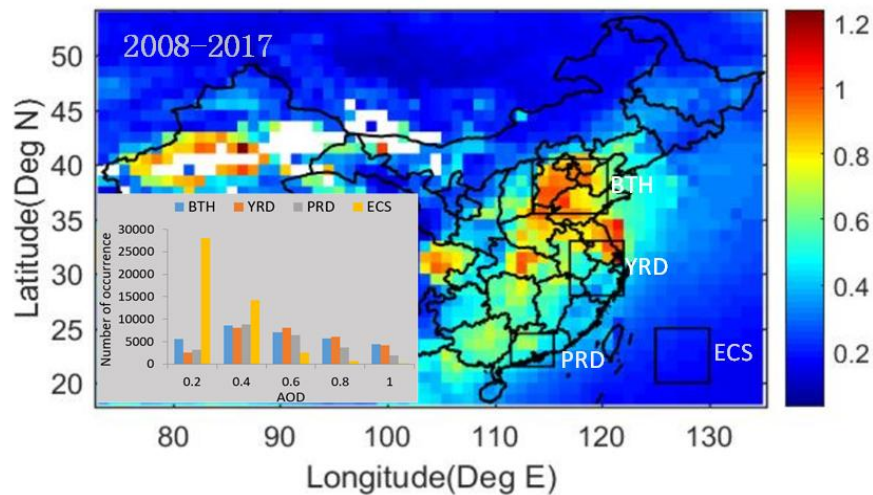
1 multi-year statistics of remotely sensed data from the two MODIS sensors aboard NASA's Terra (daytime
2 equator crossing time at 10:30 LT) and Aqua (daytime equator crossing time at 13:30 LT) satellites. The
3 retrieval of the AOD and cloud properties from both sensors allows us to explore the morning-to-afternoon
4 variation of cloud properties in conditions with either low or high AOD, over land and over ocean, and for
5 different climate regimes. This variety of conditions allows us to identify similarities and differences in the
6 effects of aerosols on clouds and thus better understand aerosol-cloud interaction. We also explore the effect
7 of meteorological history on the interaction between aerosols and clouds. We focus on low-level water
8 clouds. The paper is organized as follows. The data and region of interest are described in Section 2. The
9 main methodology is introduced in Section 3. The results and analysis are presented in Section 4. Overall
10 conclusions and potential future improvements are discussed in Section 5.

11 **2 Approach**

12 **2.1 Study area**

13 Aerosol concentrations in Eastern China are very high due to both direct emissions and secondary aerosol
14 formation from precursor gases such as NO₂, SO₂ and VOCs. They are produced by anthropogenic activities
15 such as industry, transportation and heating, black carbon and other carbonaceous aerosols produced by
16 biomass burning, dust aerosols produced from the deserts, etc. Aerosol particles influence the local climate
17 such as monsoon intensity and the distribution of precipitation. In eastern China, the monsoon in turn plays
18 an important role in the wet deposition and transport of aerosol particles (Li et al., 2016). The Asian
19 monsoon system plays an important role in the precipitation across the country (Kourtidis et al., 2015). In
20 early April, the pre-monsoonal rain period starts over southern China and the summer monsoon rain belt
21 moves northward to the Yangtze River basin in June. Further, the rain belt arrives in northern China in July
22 and the monsoon rain belt propagates back to southern China in August. The length of the rain season differs
23 between southern and northern China with the migration of the monsoon across China (Song et al., 2011).
24 Based on these characteristics, four regions with different aerosol emission levels and climate
25 characteristics were selected to study the indirect effects of aerosol particles on cloud micro- and
26 macro-physical properties. The Beijing-Tianjin-Hebei (BTH), Yangtze River Delta (YRD) and Pearl River
27 Delta (PRD) urban clusters are characterized as a temperate monsoon climate region, a subtropical monsoon
28 climate region, and a tropical monsoon climate region, respectively. The BTH domain (35.5 °N-40.5 °N,
29 113.5 °E-120.5 °E) is an area with high AOD levels due to rapid industrial and economic development (Fig.
30 1). The YRD domain (28 °N-33 °N, 117 °E-122 °E) is a major source region of black carbon (Streets et al.,
31 2001; Bond et al., 2004) and sulfate (Lu et al., 2010). The PRD domain (21.5 °N-24.5 °N, 111.5 °E-115.5 °E)
32 is an area within the intertropical convergence zone (ITCZ) migration belt, with high anthropogenic aerosol
33 emissions (Streets et al., 2003; Streets et al., 2008; Lei et al., 2011). In addition, one domain (20 °N-25 °N
34 and 125 °E-130 °E), which is located in the Eastern China Sea (ECS for short), has been selected as study
35 area for comparison. The ECS domain is relatively clean, but it is often impacted by aerosol particles
36 transported from the highly industrialized eastern China (Wang et al., 2014). The study period is 10 years,
37 i.e. 2008-2017.

1



2

3 Figure 1. Map of MODIS/AQUA level 3 AOD over Eastern China averaged over the period from 2008 to 2017. The
4 location of the four clusters (three urban and one ocean) studied here (Beijing-Tianjin-Hebei: BTH, Yangtze River Delta:
5 YRD, Pearl River Delta: PRD and Eastern China Sea: ECS) are marked with black rectangles. The inset shows a
6 histogram for the occurrence of AOD values in each of the four clusters during the period 2008-2017.

7

2.2 Data used

8

9 The aerosol and cloud properties used in this study were derived from the MODIS instruments on the Terra
10 and Aqua satellites. Since these instruments are of the same design, errors due to instrument differences are
11 minimal although some differences have been reported due to degradation of MODIS/Terra (Xiong et al.,
12 2008; Levy et al., 2010). The MODIS L3 collection 6.1 data (which was downloaded from
13 <https://ladsweb.modaps.eosdis.nasa.gov/>) provides daily aerosol and cloud parameters on a 1° by 1° spatial
14 grid. The time difference between the Terra and Aqua overpasses is about three hours, with variations due to
15 swath width. In the following, the time difference between the MODIS/Terra and Aqua observations is
16 referred to as the timestep. The application of daily MODIS satellite data on a 1° by 1° spatial grid in this
17 study on aerosol-cloud interaction (ACI) ensures that the aerosol and cloud retrievals are coincident. The
18 MODIS instruments have 36 spectral bands, the first seven of these (0.47- 2.13 μm) are used for the
19 retrieval of aerosol properties (Remer et al., 2005) while cloud properties are retrieved using additional
20 wavelengths in other parts of the spectrum (Platnick et al., 2003). More detailed information on algorithms
21 for the retrieval of aerosol and cloud properties is provided at <http://modis-atmos.gsfc.nasa.gov>. In this
22 study on ACI we use the AOD at 550 nm (referred to as AOD throughout this manuscript), CDR, cloud
23 liquid water path (CWP), cloud optical thickness (COT), cloud fraction (CF), cloud top pressure (CTP) and
24 cloud top temperature (CTT) from both instruments. AOD is used as a proxy for the amount of aerosol
25 particles in the atmospheric column to investigate ACI (Andreae, 2009; Kourtidis et al., 2015). To reduce a
26 possible over-estimation of the AOD, cases with AOD greater than 0.8 were excluded from further analysis.
27 The focus of this study is on warm clouds with CTP larger than 700 hPa, CTT larger than 273K and CWP
28 lower than 200 g m⁻², as most aerosols exist in the lower troposphere (Michibata et al., 2014).

29

30 In addition, to explore the effect of meteorological conditions on ACI, we use the daily temperature at the
31 1000 hPa and 700 hPa levels, relative humidity at the 750hPa level and pressure vertical velocity (PVV) at
the 750 hPa level. LTS is defined as the difference in potential temperature between the free troposphere
(700hpa) and the surface, which can be regarded as a measure of the strength of the inversion that caps the

1 planetary boundary layer (Klein and Hartmann, 1993; Wood and Bretherton, 2006). These meteorological
2 data were obtained from daily ERA Interim Reanalysis data which contains global meteorological
3 conditions on a grid of $1^\circ \times 1^\circ$ with 37 levels in the vertical (1000-1 hPa) every six hours (00:00, 06:00, 12:00,
4 18:00 UTC) (<http://apps.ecmwf.int/datasets/data/interim-full-daily/>). The meteorological properties were
5 resampled to 10:30 (local time) by taking a weighted average of the properties at the two closest times
6 (00:00 UTC and 06:00 UTC) provided by ERA Interim.

7 In this study, high and low AOD are defined as the highest and lowest quartile for each $1^\circ \times 1^\circ$ location to
8 reduce climatological spatial gradients in aerosol and cloud parameters. As a result, the difference between
9 high and low AOD varies by location. So, for each $1^\circ \times 1^\circ$ grid cell, 3642 data samples are available for the
10 10-year study period.

11 **3 Method**

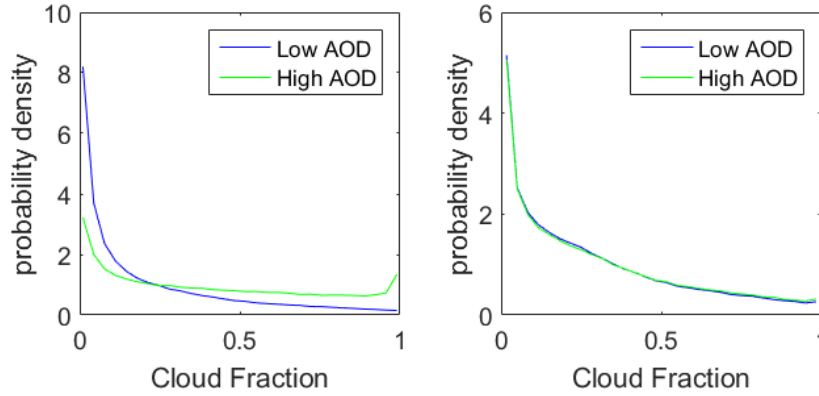
12 **3.1 Normalization for initial background**

13 For the comparison of the difference in cloud properties in high and in low AOD conditions and the change
14 in this difference during the time step, we need to ensure that the initial conditions are similar, i.e. the
15 probability distributions of a cloud parameter Cloud_X at the start of the time step for the low and high AOD
16 cases should be similar. Any change in this distribution at the end of the time step can then be attributed to
17 changes in cloud properties due to aerosol and/or meteorological effects. To reduce the difference between
18 the initial probability distribution of Cloud_X in high and low AOD conditions at the start of the timestep,
19 normalized histograms of cloud properties and meteorological parameters are made for high and low AOD
20 conditions following the method described by Gryspeerdt et al. (2014).

21 In Fig. 2 we illustrate the process to remove possible effects linking, as an example, CF and AOD.
22 Normalized histograms of CF are made for the high and low AOD conditions following Gryspeerdt et al.
23 (2014) with the difference that in the current study AOD is used instead of AI (Andreae, 2009; Kourtidis, et
24 al., 2015). The CF probability density functions for low and high AOD conditions at the start time are
25 different as illustrated in Fig. 2a. This difference indicates a link between CF and AOD at the start of the
26 time step which needs to be removed to detect the effect of changes during the time step. This is achieved,
27 following the process described in more detail by Gryspeerdt et al. (2014). In brief, for each bin data points
28 are drawn out randomly from the conditions with the larger probability density frequency until both
29 distributions match. This is performed independently for each bin and the entire process is repeated until the
30 normalised histograms in both AOD conditions are similar. As a result of this normalization process, the CF
31 distributions at the start of the timestep are nearly identical for both AOD conditions, i.e. the non-aerosol
32 effect linking CF and AOD has been removed. This technique has also been applied to ensure that the high
33 and low AOD conditions have the same probability distributions for CDR, COT, CWP and CTP at the start
34 time. Among those cloud properties, this process of normalization has the greatest effect on the cloud
35 fraction and its dependence on aerosol-cloud interaction. Throughout the work, we only take a subset of
36 original data by removing random samples until the histograms are similar.

37 Note that here and in the following sections, normalised histograms of cloud properties for the high and low
38 AOD populations are made for the whole region (Section 3.1), because the data volume based on each $1^\circ \times$
39 1° location is relatively small. However, the difference between the cloud properties for low and high AOD

1 at the start time is based on each $1^\circ \times 1^\circ$ location (Section 4.1). So the difference of the cloud properties
 2 between the low and high AOD at the start time is not zero.



3
 4 Figure 2 An example of the probability density distribution of warm cloud fraction (CF) for low and high AOD
 5 conditions. (a) there is a strong link between AOD and CF before histogram normalization, (b) the link is reduced after
 6 histogram normalization.

7 3.2 The definition of $d(\text{Cloud_X})$

8 After removal of the potential relationships between AOD and cloud parameters at the time of the Terra
 9 (morning) overpass, as described in Sect. 3.1, effects of aerosol particles on cloud properties are investigated
 10 from the change in the relationship between AOD and cloud parameters over the timestep. For cloud property
 11 Cloud_X (where X = CF, COT, CWP, CDR or CTP), the change during the timestep is indicated by
 12 $\Delta\text{Cloud_X}$. The mean $\Delta\text{Cloud_X}$ for high AOD is then indicated by $\overline{\Delta\text{Cloud_X}[\text{High AOD}]}$ and similar for
 13 low AOD. The difference between the mean change in Cloud_X during the timestep in high and low AOD
 14 conditions is then indicated by $d(\text{Cloud_X})$:

$$15 \quad d(\text{Cloud_X}) = \overline{\Delta\text{Cloud_X}[\text{High AOD}]} - \overline{\Delta\text{Cloud_X}[\text{Low AOD}]}$$

16 The high AOD is representative of polluted atmospheric conditions, and the low AOD is representative of
 17 clean atmospheric conditions. The difference ($d(\text{Cloud_X})$) between the mean values of the cloud property
 18 Cloud_X during clean (low AOD) and polluted (high AOD) conditions indicates the effect of these two
 19 aerosol cases on the cloud property Cloud_X. For example, $d(\text{CWP})$ would be the difference between the
 20 mean change in CWP in high AOD conditions minus that in low AOD conditions.

21 Student's t test is used to determine whether two data sets are significantly different from each other. The
 22 marker ** at the top right corner of symbol "+" (or "-") denotes that the difference between a change in cloud
 23 property and zero is significant (at 95% confidence level).

24 4 Results and Discussion

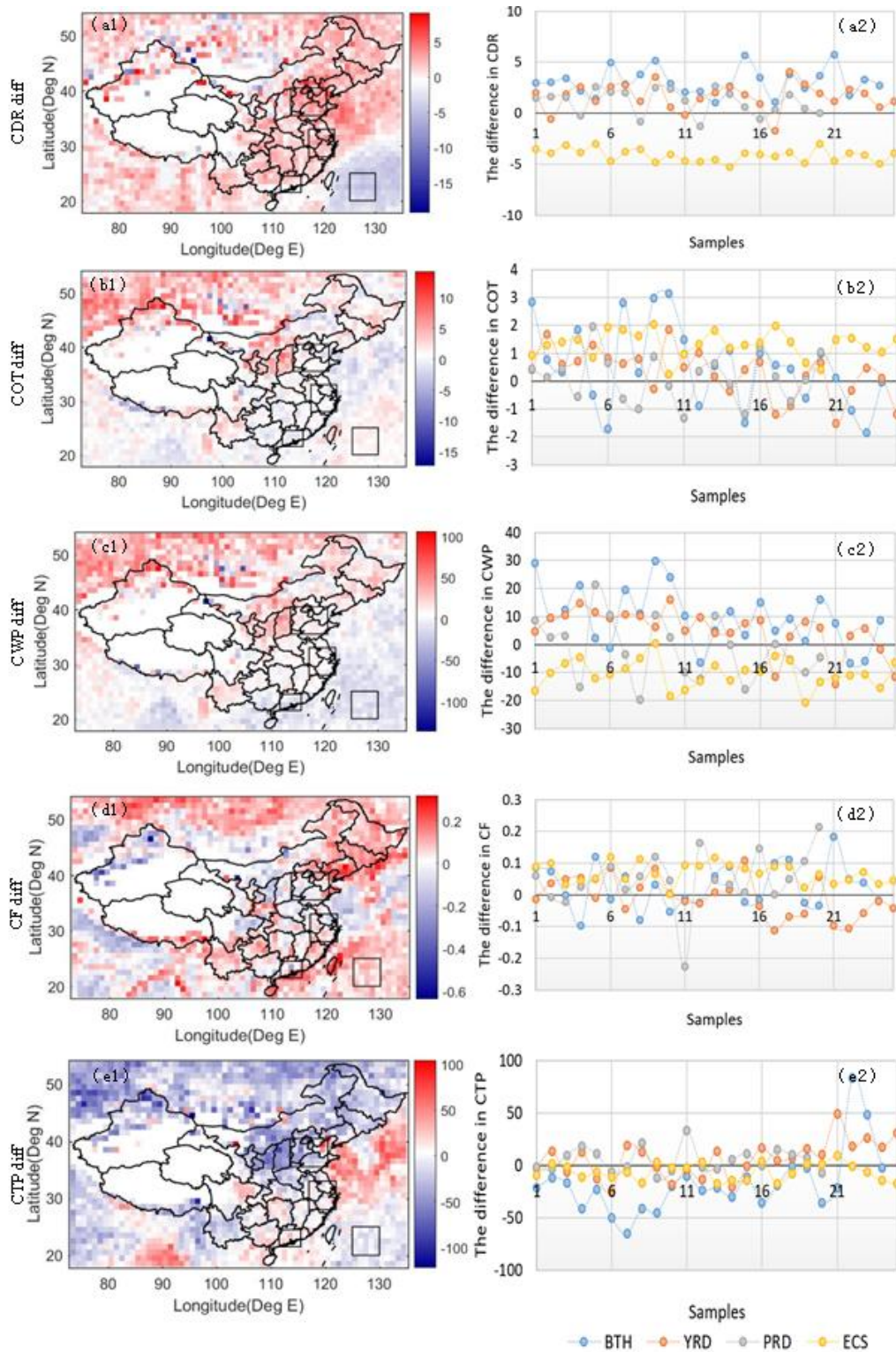
25 4.1 The difference of cloud properties between the low and high AOD at the start time

26 The difference in the mean cloud properties (CDR, CF, COT, CWP and CTP) during high and low AOD
 27 conditions at the start time for each $1^\circ \times 1^\circ$ grid cell, i.e.,

28 $\left\{ \overline{\text{Cloud_X}[\text{High AOD}]} - \overline{\text{Cloud_X}[\text{Low AOD}]} \right\}_{t=0}$ represents the change in cloud properties due to the

29 higher AOD. Figure 3 shows the spatial distributions of these differences (left column) and sample series of

1 the difference (right column) for the four regions of interest. The selection of samples for each region is
2 according to the pixels in the region. Figures 3(a1-a2) show that over the ECS, CDR is smaller at high AOD
3 than at low AOD, which is consistent with Twomey's effect. In contrast, over the three urban clusters, CDR is
4 larger at high AOD. This behavior has been observed before for warm clouds in conditions with high AOD
5 (Liu et al., 2017) and may result from the intense competition for the available water vapour and the
6 evaporation of smaller droplets as a consequence of the high aerosol abundance over these regions (Yuan et
7 al., 2008; Tang et al., 2014; Wang et al., 2014; Liu et al., 2017). For COT (figures 3(b1-b2)) the values are
8 significantly higher at high AOD over the ECS and the BTH, however, COT does not show a significant
9 difference between the situations at low and high AOD over the YRD and PRD. These results indicate that
10 there is no clear dependence of COT on aerosol load, and also the aerosol type may influence the aerosol
11 effect on COT. Figures 3(c1-c2) show that CWP is lower at high AOD over the ECS, which is in clear
12 contrast with the so-called "lifetime effect" proposed by Albrecht in 1989. In contrast, over the BTH, CWP
13 behaves similar to COT and is higher at high AOD. Furthermore, CWP is also higher at high AOD.
14 Ackerman et al. (2004) reported that CWP is not generally observed larger, but significantly smaller in high
15 AOD conditions. They reported that CWP response to the increasing AOD is determined by the balance of
16 two competitive factors: moistening from precipitation decrease and drying from increasing entrainment of
17 dry overlaying air. With increasing AOD, CF does not show any significant correlation between changes in
18 AOD and CDR variations over the BTH and YRD. However, CF is larger at high AOD over the PRD and
19 ECS. Wang et al., (2014) also found that when aerosol loading is relatively small, cloud cover is found to
20 increase over the YRD and ECS in response to aerosol enhancement regardless of RH conditions. Meanwhile,
21 over the YRD urban cluster CTP is higher at high AOD, as suggested by Liu et al. (2017). In contrast, CTP is
22 lower at high AOD over the BTH and ECS. Many studies have also reported that with higher cloud altitude
23 CTP decreases in most of the places as AOD increases except for some regions at low AOD (Myhre et al.,
24 2007; Kaufman et al., 2005 and Alam et al., 2010). This might have resulted from the suppression of
25 precipitation by increasing cloud lifetime and thus also affecting the cloud albedo and cloud top pressure.
26



1

2 Figure 3 Spatial distribution of the differences in cloud properties (top to bottom: CDR, COT, CWP, CF and CTP)
 3 between the highest and the lowest MODIS AOD quartiles (highest - lowest) at the start time of the timestep
 4 (MODIS/Terra) (left, a1-e1) and sample series of the differences in cloud properties (CDR, COT, CWP, CF and CTP)
 5 between the highest and the lowest MODIS AOD quartiles (highest - lowest) at the start time of the timestep
 6 (MODIS/Terra) (right, a2-e2) over Eastern China for the time period 2008-2017. See legend at the bottom for the
 7 meaning of the colours identifying the different regions.

8 To better characterize the variation in cloud properties between high and low AOD, Table 1 summarizes the
 9 difference in cloud properties between high and low AOD at start time for the four study areas. We find that
 10 different regions with various aerosol emission levels and different climate characteristics show different

1 ACI patterns. Some links between aerosol and cloud in the four regions are different from those of previous
 2 studies over China (Wang et al., 2014; Tang et al., 2014; Kourtidis et al., 2015; Liu et al., 2017), which
 3 might be due to the use of different data sets (MODIS C6.1 versus older versions), hypothesis and target
 4 areas characterized by complex aerosol composition and varying meteorological conditions. Overall, the
 5 result implies that the interaction between aerosol particles and clouds is more complex and of greater
 6 uncertainty over land (BTH, YRD and PRD) than over ocean (ECS). Jin and Shepherd (2008) also noted that
 7 aerosols affect clouds more significantly over ocean than over land. They suggested that dynamic processes
 8 related to factors like urban land cover may play at least an equally critical role in cloud formation.

9 Table 1 The responses of cloud properties to the increasing AOD

Parameters	AOD	CDR	COT	CWP	CF	CTP
BTH	+	+**	+**	+**	+	-**
YRD	+	+**	+	+**	-	+**
PRD	+	+**	+	-	+**	+**
ECS	+	-**	+**	-**	+**	-**

10 Note: '+' indicates increasing, '-' indicates decreasing and '**' at the top right corner of the symbol '+' (or '-') denotes
 11 that the difference between a change in cloud property and zero is significant (at 95% confidence level).

12 4.2 The meteorology of the four target regions

13 The meteorological and aerosol effects on clouds are reported to be tightly connected, and this connection
 14 must be accounted for in any study of aerosol-cloud interactions (Stevens and Feingold, 2009; Koren et al.,
 15 2010a). Although normalized histograms of meteorological parameters are made for high and low AOD
 16 conditions at the start time, the normalization described in Sect. 3.1 is based on the whole region.
 17 Differences in meteorological conditions may still occur between each 1° by 1° grid cell. In this study,
 18 we analyze the meteorology of the different regions, in support of the interpretation of the regional variation
 19 of the relationships between aerosols and clouds.

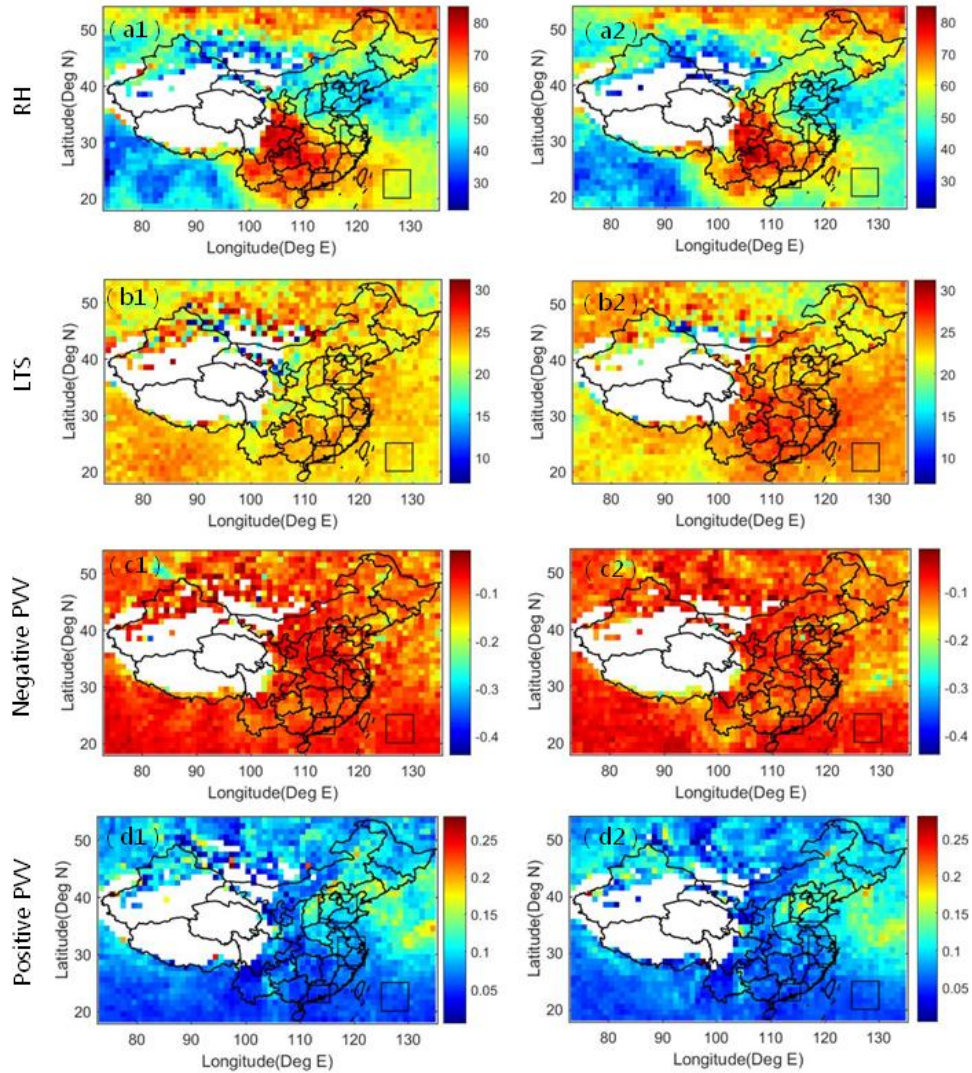


Figure 4 Spatial distributions of meteorological parameters (top to bottom: RH, LTS, positive PVV and negative PVV) at the start time of the timestep (MODIS/Terra) for low AOD conditions (left, a1-d1) and for high AOD conditions (right, a2-d2). All the data are averaged over all years between 2008 and 2017.

The spatial variations of the aerosol and cloud properties over the four regions, averaged over the years 2008-2017, are shown in Fig. 4. Over the urban clusters, we can see an increasing north–south pattern in RH and LTS, with the lowest values found in the PRD. For the negative PVV, the spatial distributions for the low and high AOD situations are remarkably similar, with the highest values over the BTH and decreasing toward the south to near zero over the PRD. In contrast, the positive PVV is smallest over the BTH, with little variation over the study area. Overall, the meteorological parameters over the YRD and PRD are similar to those over the ECS, irrespective of the AOD. Furthermore, the LTS is significantly larger in the high AOD conditions for all four regions. Zhao et al. (2006) proposed that the enhancement in atmospheric stability tends to depress upward motion and precipitation, leading to an increase in aerosol particles. The spatial distributions of both positive and negative PVV in the low AOD conditions are similar to those in high AOD conditions.

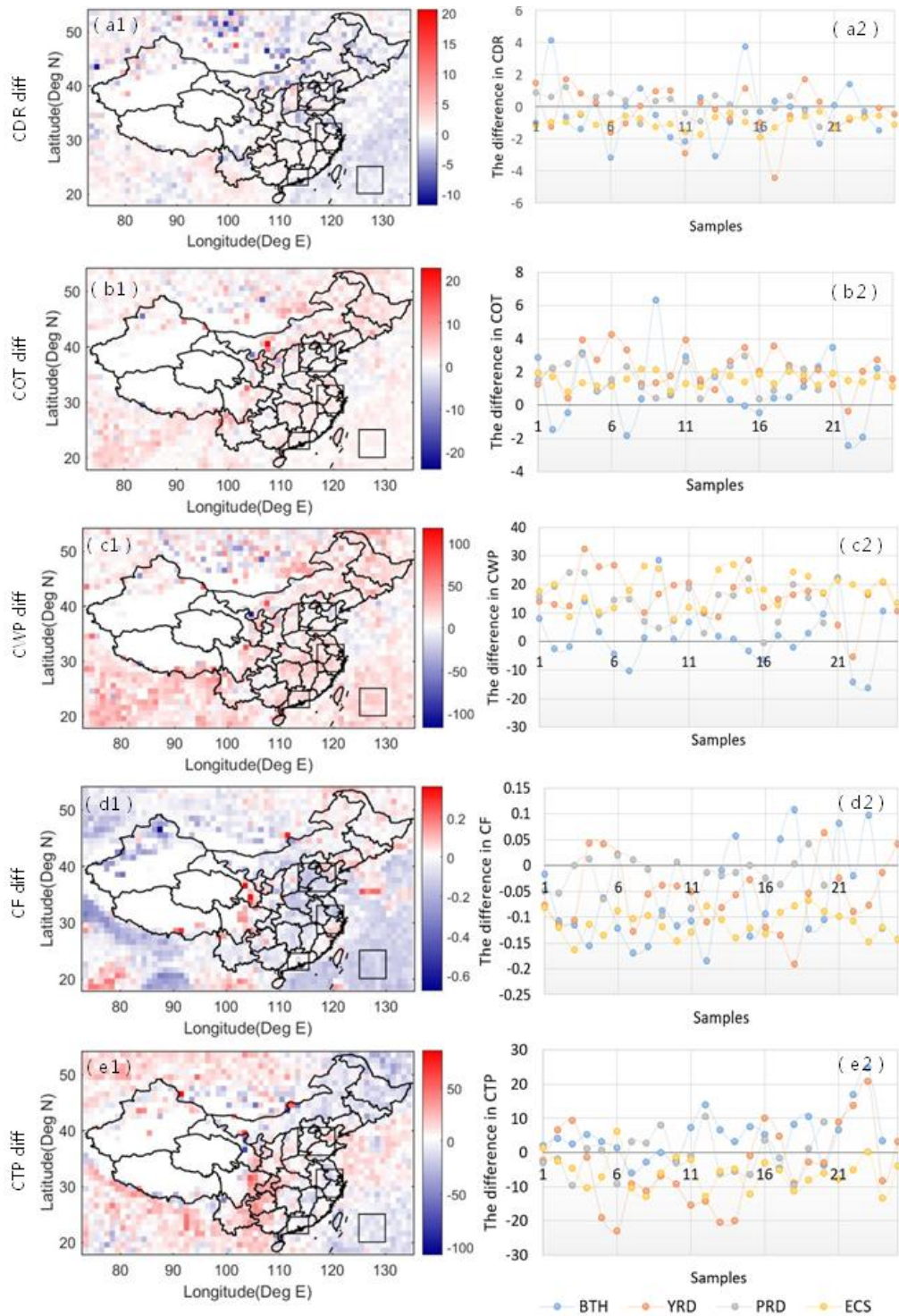
4.3 The mean change in cloud properties over the timestep for low and high AOD

The differences between the mean afternoon and morning values of cloud properties in each $1^\circ \times 1^\circ$ grid cell in either low or high AOD conditions shows the variation of cloud properties during 3 hours of cloud evolution

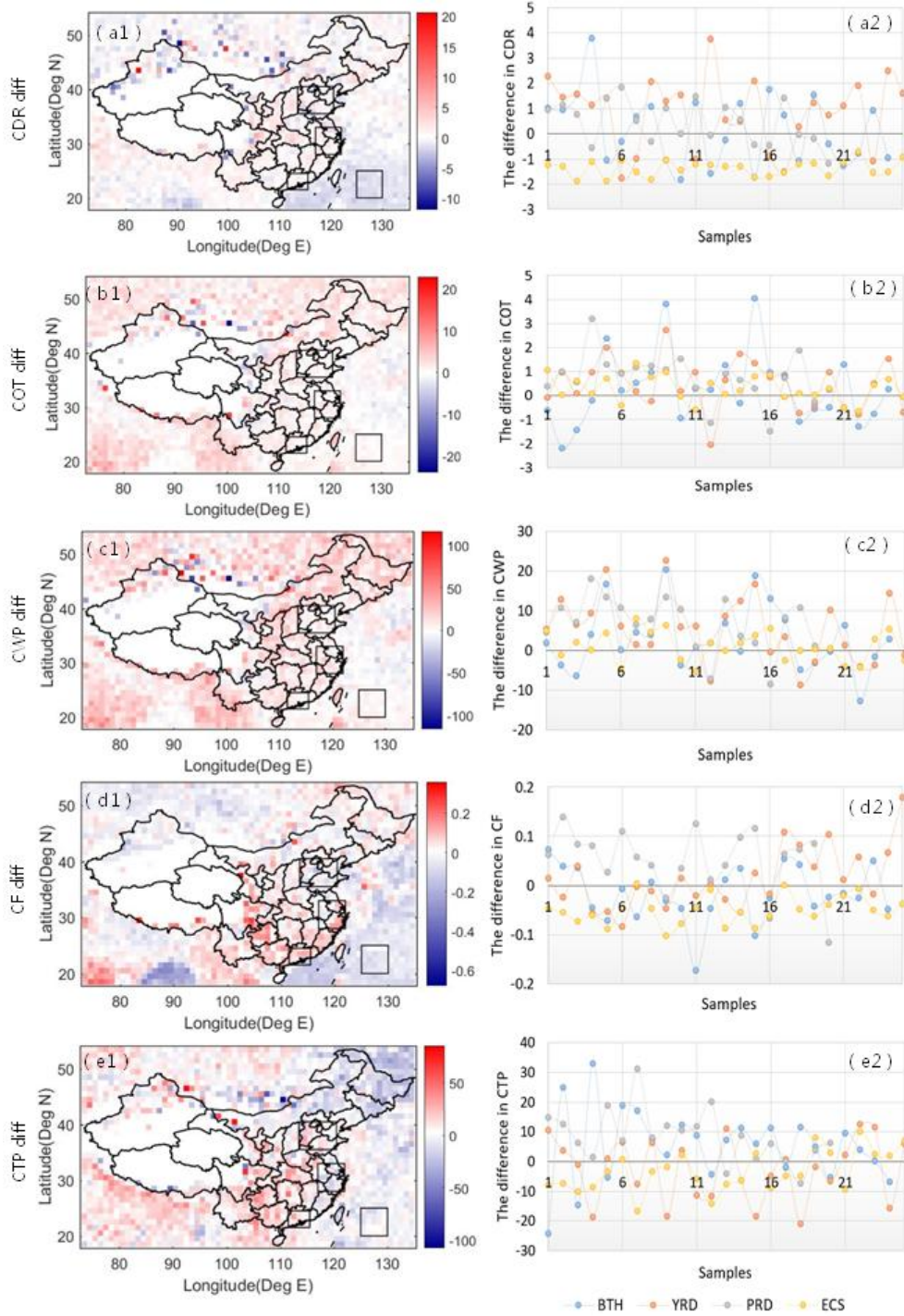
1 at low/high aerosol concentrations. Figure 5 presents the spatial distributions (left, a1-e1) and the sample
2 series (right, a2-e2) of differences in cloud properties (CDR, COT, CWP, CF and CTP) after this 3-hour
3 period for the lowest MODIS/Terra AOD quartiles. Figure 6 shows the spatial distributions (left, a1-e1) and
4 sample series (right, a2-e2) of these differences for the highest MODIS/Terra AOD quartiles.

5 Overall, we look at statistics for a large dataset of 10 years. Concerning the effect of aerosol loading on
6 cloud parameters in each urban cluster, a decrease of CF occurs over the BTH for low AOD conditions,
7 which is opposite to the CTP variation for both AOD conditions. For the variations of CDR over the YRD
8 urban cluster, a significant increase occurs under high AOD conditions, which may be attributed to the
9 higher RH (see figure 4(a1, a2)). As regards the variation of CF and CTP, a significant decrease occurs
10 under low AOD conditions. Likewise, an increase of the CDR was observed for high AOD conditions over
11 the PRD urban cluster. Furthermore, decreases of CF and CTP were observed for low AOD conditions and
12 increases of CF and CTP were observed for high AOD conditions. From the perspective of considering all
13 urban clusters (BTH, YRD and PRD), both COT and CWP increase over land during the 3 hours timestep
14 for both low and high AOD. Overall, the variation in cloud properties after the timestep over BTH is less
15 significant than over the YRD and PRD for both low and high AOD conditions. This may result from less
16 humid and more unstable atmospheric environments over the BTH than over the other two urban clusters
17 (as shown in Section 4.2). Over the ECS, in both low and high AOD conditions, CDR, CF and CTP
18 decrease during the timestep while COT and CWP increase (see Figure 5).

19 In general, the variations over 3 hours in COT and CWP over land are similar to those over ocean for both
20 low and high AOD conditions. Another similarity is that CF decreases for low AOD conditions over land
21 and ocean during the 3h timestep. Having a closer look at the CF variation over the YRD and PRD, we see
22 that CF increases in high AOD conditions during the 3h timestep. This implies that the variation of CF
23 may depend on the initial AOD conditions. The decrease in afternoon cloud cover over ocean confirms that
24 the largest cover for marine clouds is reached early in the morning as was also concluded by Meskhidze et al.
25 (2009). Meanwhile, a significant difference is found between land and ocean areas, i.e. in high AOD
26 conditions CDR increases over land but decreases over ocean during the 3h timestep. Table 2 summaries the
27 differences in cloud properties between the Aqua and Terra overpasses for high and low AOD conditions
28 over land and ocean during the time period 2008-2017.



1
2 Figure 5. Spatial distributions of differences in cloud properties (CDR, COT, CWP, CF and CTP) between Aqua and
3 Terra overpasses (3 hours) for the lowest MODIS/Terra AOD quartiles (left, a1-e1). Sample series of the differences in
4 cloud properties (CDR, COT, CWP, CF and CTP) between the values at the start time and the end time of the timestep
5 for the lowest MODIS AOD quartiles (right, a2-e2).



1
2 Figure 6. Spatial distributions of differences in cloud properties (CDR, COT, CWP, CF and CTP) between Aqua and
3 Terra overpasses (3 hours) for the highest MODIS/Terra AOD quartiles (left, a1-e1). Sample series of the differences in
4 cloud properties (CDR, COT, CWP, CF and CTP) between the values at the start time and the end time of the timestep
5 for the highest MODIS AOD quartiles (right, a2-e2).

6 Table 2 Differences in cloud properties between Aqua and Terra for high and low AOD, over land and ocean.

Parameters	CDR	COT	CWP	CF	CTP
BTH L_AOD	-	+**	+	._**	+**
BTH H_AOD	+	+	+**	-	+**
BTH d(Cloud_X)	-	-	-	._**	+**
YRD L_AOD	-	+**	+**	._**	._**
YRD H_AOD	+**	+**	+**	+	-
YRD d(Cloud_X)	+	._**	._**	._**	-
PRD L_AOD	+	+**	+**	._**	._**

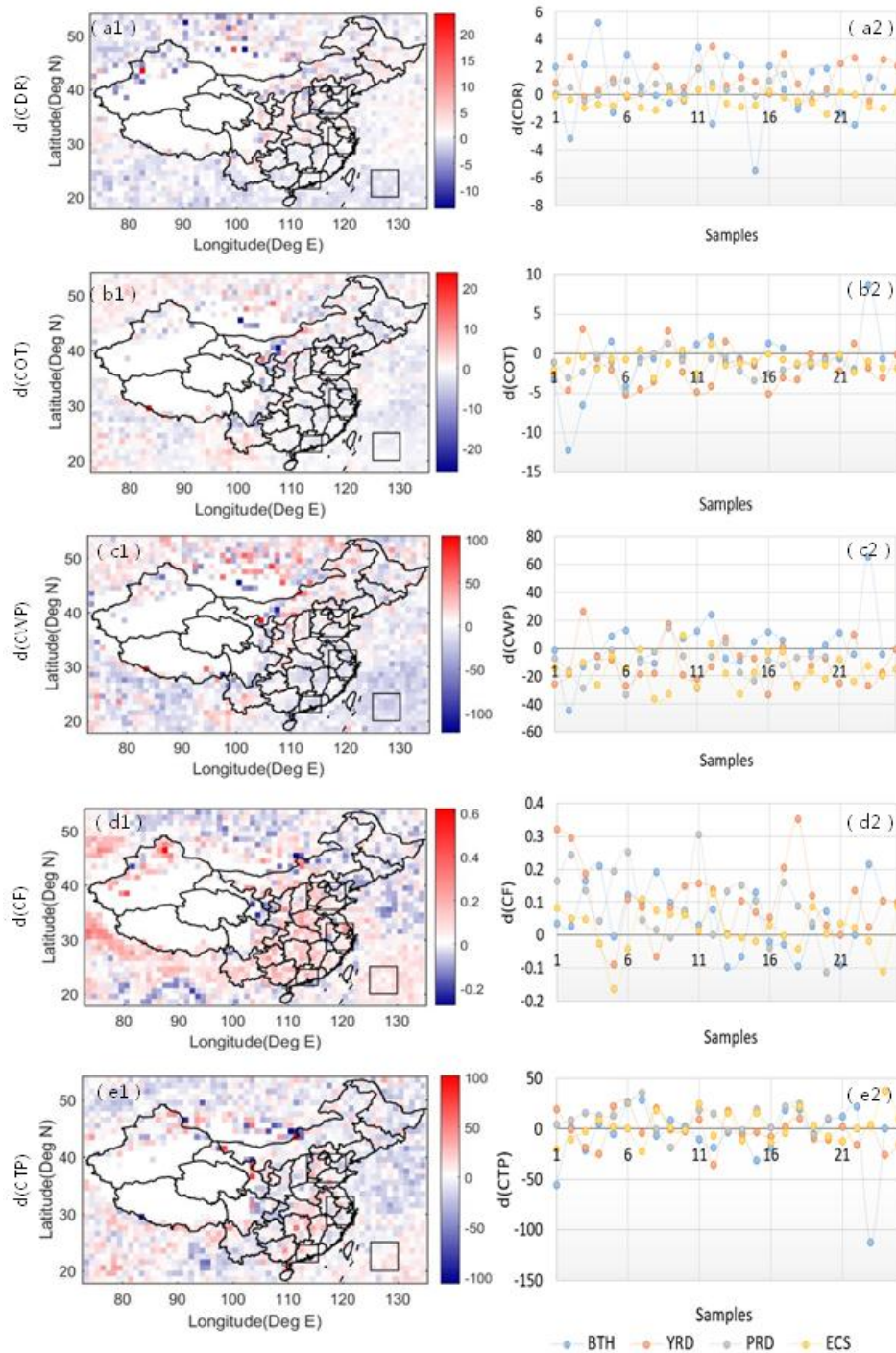
	H_AOD	***	***	***	***	***
	d(Cloud_X)	+	***	***	***	***
	L_AOD	***	***	***	***	***
ECS	H_AOD	***	***	+	***	***
	d(Cloud_X)	***	***	***	***	-

Note: '+' indicates increasing, '-' indicates decreasing and ** at the top right corner of symbol "+" (or "-") denotes that the difference between a change in cloud property and zero is significant (at 95% confidence level).

The differences between the mean changes in cloud properties (CF, COT, CWP, CDR and CTP) between the Terra and Aqua overpasses in high and in low AOD conditions (d(Cloud_X) as defined in Section 3.2) are investigated to identify the effect of aerosol particles on the cloud properties. Figure 7 shows the differences between the mean change in cloud properties at low and high AOD conditions during the two observations at 10:30 and 13:30.

Figure 7 shows that the values of d(CDR) over the three urban clusters are not mostly positive or negative, which indicates that in high AOD conditions over land the variation in CDR during the three hours between the MODIS/Terra and Aqua overpasses is similar. Over the ECS the values of d(CDR) are positive, which indicates that the CDR in high AOD conditions decreases much more than during low AOD conditions over ocean. Wang et al. (2014) also reported a negative correlation between CDR and AOD over the ECS, in accordance with the Twomey effect. Furthermore, CDR tends to be smallest in polluted and strong-inversion environments, an outcome in good agreement with the findings of Matsui et al. (2006). Most of the d(COT) values are negative over the four regions, especially for the YRD, PRD and ECS. This shows that the COT increases less in high AOD conditions than in low AOD conditions, over both land and ocean, which is contrast with the findings of Meskhidze et al. (2009). Likewise, the values of d(CWP) are almost all negative over the four regions although over the BTH urban cluster the values are not clear. This indicates that in high AOD conditions the CWP increases less during the timestep than in low AOD conditions, a result in accordance with the conclusion that higher LTS is linked with a slightly lower CWP (Matsui et al., 2006). We can conclude that the variation trend of COT and CWP after 3 hours depends little on the initial AOD, but the initial AOD conditions can affect the amplitude of variation of COT and CWP. Meanwhile, the values of d(CF) are smaller than zero over the ECS. This shows that the cloud fraction in high AOD conditions over the ECS decreases less than that in low AOD conditions. However, Meskhidze et al. (2009) found that an increase of the aerosol concentration may lead to enhanced reduction of afternoon cloud coverage and optical thickness for marine stratocumulus regions off the coast of California, Peru, and southern Africa. Therefore, the connection between AOD and variation of cloud cover could be a response to regional-scale changes in aerosol covarying with meteorological conditions. The value of d(CF) is overall positive over the PRD, which indicates that over the PRD in high AOD conditions the cloud cover increases much more than the cloud cover decreases in low AOD conditions. Mauger and Norris (2007) have shown that scenes with large AOD and large cloud fraction experienced greater LTS. As regards CTP, we find that the values of d(CTP) are positive over the BTH and PRD urban cluster, but the values of d(CTP) over the other two regions do not show a clear pattern. This indicates that in high AOD conditions over the PRD region the CTP increases much more than the CTP decreases in low AOD conditions. We can conclude that the variation in d(Cloud_X) is different for continental and oceanic clouds. This applies to CDR, cloud fraction (CF) and CTP, but not to COT and CWP. Table 2 summarizes the differences between the mean changes in cloud properties for low and high

1 AOD over the timestep of 3 hours.
 2 Based on the above findings, we conclude that over the ECS the values of CDR, CWP and CTP are smaller
 3 but the values of COT and CF are larger in high AOD conditions. After the 3 hours timestep, CDR, CF
 4 and CTP become smaller, irrespective of the AOD. Furthermore, CDR decreases much more in high AOD
 5 conditions but CF and CTP decreases much more in high AOD conditions. In contrast, COT and CWP
 6 become larger in both AOD conditions, more significantly in low AOD conditions. Over the urban
 7 clusters, COT and CWP also increase over the timestep in both AOD conditions, especially for the low
 8 AOD condition. For CF the values in low AOD conditions decrease over the timestep. The CTP change
 9 behaves different among the three urban clusters during the 3 hours.



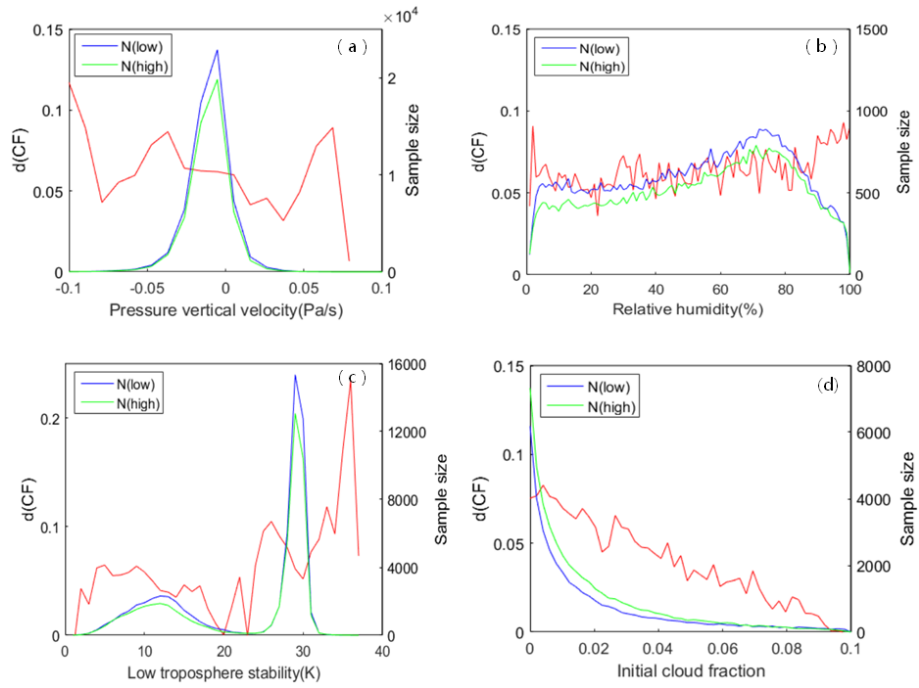
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 11 Figure 7 Spatial distributions (left, a1-e1) and sample time series (right, a2-e2) of d(Cloud_X) (as defined in sect.
 12 3.2) for CDR, COT, CWP, CF and CTP over Eastern China during the time period 2008-2017.

1 4.4 Meteorological effects

2 In order to explore the initial meteorological effects on the correlations between AOD and the cloud
3 fraction, we determine the difference in mean cloud parameters between the high and low AOD
4 conditions at the end of the timestep ($d(\text{Cloud}_X)$) in meteorological variable space rather than in
5 longitude-latitude space. Therefore, we define high and low AOD as the highest and lowest quartile
6 for each bin of the meteorological parameters, respectively. Figure 8 shows the effect of meteorological
7 factors (PVV, RH, LTS and initial cloud fraction) on the $d(\text{CF})$ when the cloud cover increases
8 ($\Delta\text{Cloud}_X > 0$) under both low and high AOD conditions over land after the 3 hours timestep. Figure 9
9 shows the effect of meteorological factors on the $d(\text{CF})$ when the cloud cover decreases ($\Delta\text{Cloud}_X < 0$)
10 under both low and high AOD conditions over land after the 3 hours timestep. From both figures we find
11 that almost all $d(\text{CF})$ values are positive, indicating that the variations of CF are larger in high AOD than
12 that in low AOD after the 3 hours timestep..

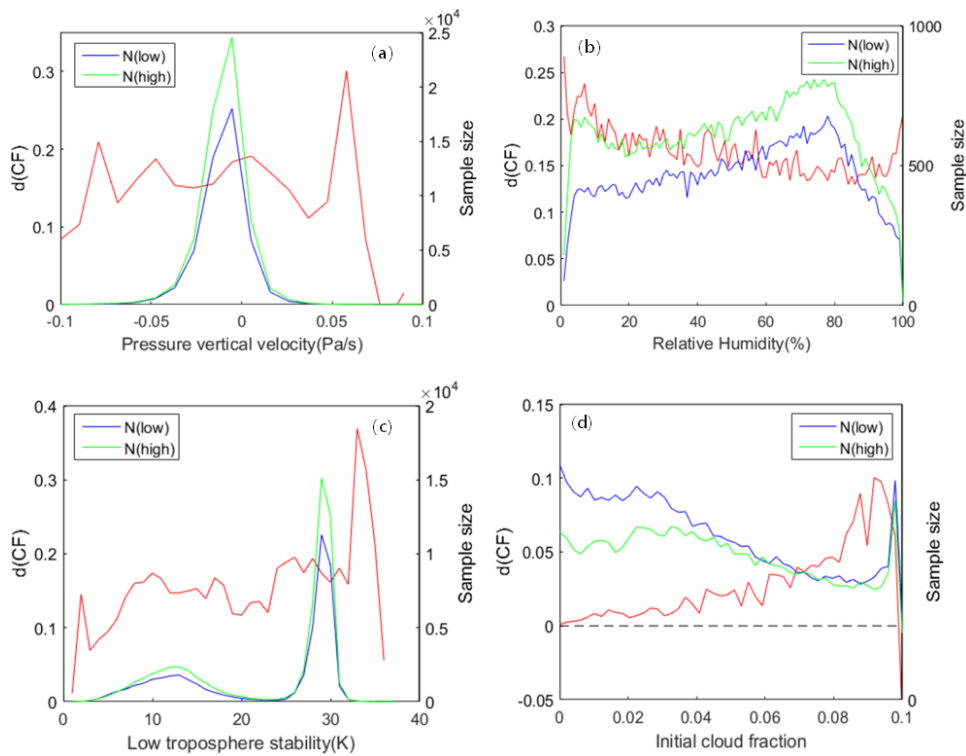
13 The PVV, a measure of dynamic convection strength, is very important for cloud formation. Negative
14 PVV is indicative of upward air motion, adiabatic expansion and cooling and hence, if cooling is
15 sufficient, cloud formation (Jones et al., 2009). Figure 8(a) shows that the $d(\text{CF})$ decreases with the PVV
16 over the range from -0.05 Pa s^{-1} to 0.05 Pa s^{-1} as cloud cover increases in both conditions over the
17 3 hours timestep. This indicates that the weaker downward motion and stronger upward motion of air
18 parcels makes the difference between the increment of cloud cover in high and low AOD conditions larger: In
19 other words, the increase rate of cloud cover is larger for high AOD under stronger upward motion of air
20 parcels. Jones et al. (2009) showed that stronger upward motion of air parcels can promote the cloud
21 formation in both high and low AOD conditions, but they did not report the increase rate of cloud
22 formation in both AOD conditions. While cloud cover decreases in both conditions over the 3 hours
23 timestep, Figure 9(a) shows that the $d(\text{CF})$ increases with the PVV over the range from -0.05 Pa s^{-1} to 0
24 Pa s^{-1} and decreases with the PVV over the range from 0 Pa s^{-1} to 0.05 Pa s^{-1} . This indicates that the
25 decrease rate of cloud cover is smaller for high AOD both under stronger upward motion of air parcels
26 and stronger downward motion of air parcels. Outside this range of PVV values the relationship
27 becomes harder to determine due to the reduced data volume in both cases. Figure 8(b) shows that the
28 $d(\text{CF})$ decreases with increasing RH when RH is lower than 20%. This implies that the increase rate of

1 cloud cover is smaller for high AOD with increasing RH. However, when RH is larger than 20%, the
2 increase rate of cloud cover is larger for high AOD with increasing RH. An increase of $d(\text{CF})$ occurs due
3 to activation of CCN and formation of clouds (Feingold et al., 2003; Liu et al., 2017). It should be noted
4 that the variation of $d(\text{CF})$ with increasing RH above around 80% is uncertain as the sample sizes of high
5 and low AOD conditions are small. In contrast, the $d(\text{CF})$ values become smaller with increasing RH
6 over the whole RH range (See Figure 9(b)), indicating that the decrease rate of cloud cover is smaller for
7 high AOD than that for low AOD. The LTS is an indicator for the mixing state of the atmospheric layer
8 adjacent to the surface. It describes to some extent the atmosphere's tendency to promote or suppress
9 vertical motion (Medeiros and Stevens, 2011), which in turn affects cloud properties (Klein and
10 Hartmann, 1993). Low LTS represents a relatively unstable atmosphere and high LTS represents a more
11 stable atmosphere. Both Figure 8(c) and Figure 9(c) show that the $d(\text{CF})$ increases and then decreases
12 with increasing LTS when LTS is lower than 20K, but increases with increasing LTS for higher values
13 (LTS >20K). However, the sample sizes of high and low AOD conditions are extremely disproportionate
14 when LTS is larger than 20K. Therefore, it is difficult to reach a conclusion from the relationship
15 between $d(\text{CF})$ and LTS when LTS is larger than 20K. Figure 8(d) shows a strong negative relationship
16 between $d(\text{CF})$ and initial cloud fraction. The $d(\text{CF})$ increases with increasing initial cloud cover, even
17 though the data volume becomes smaller over the range from 0 to 1.0. This implies that the increase rate
18 of cloud cover becomes smaller in high AOD environment than that in low AOD environment with an
19 increase of the initial cloud cover. Likewise, Figure 9(d) also shows that $d(\text{CF})$ decreases with increasing
20 initial cloud cover, indicating that the decrease rate of cloud cover becomes larger in high AOD
21 environment than that in low AOD environment. This phenomenon is different from the observed weak
22 relationship between $d(\text{CF})$ and initial cloud fraction in the oceanic shallow cumulus regime (Gryspeerdt
23 et al., 2014). It may result from the combination of above two cases.



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Figure 8. Variation of $d(CF)$ (red) as function of initial meteorological parameters and cloud fraction for warm clouds when the cloud cover increases under both low and high AOD conditions over land after the 3 hours timestep. The distribution of points for low (blue) and high (green) AOD as a function of meteorological parameters is shown by the solid lines. This plot is composed from MODIS data (including Terra and Aqua) for all warm cloud points over the years 2008-2017. Meteorological parameters are plotted along the horizontal axis, the left vertical axis denotes $d(CF)$ and the right vertical axis denotes the number of high and low AOD samples.



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Figure 9 The same as Fig. 8 but for warm clouds when the cloud cover decreases under both low and high AOD conditions over land after the 3 hours timestep.

1 5 Conclusions

2 The large anthropogenic emissions in eastern China render this area an important hotspot for studying
3 how cloud microphysical properties are affected by anthropogenic aerosols (Ding et al., 2013). In this
4 work, based on the near-simultaneous aerosol and cloud retrievals provided by MODIS, together with the
5 ERA Interim Reanalysis data, we investigated the effect of aerosol loading, using AOD as a proxy, on
6 aerosol-cloud interactions. Aerosol-cloud interaction was studied over three major urban clusters in
7 eastern China and over one area over the Eastern China Sea. These four areas are representative of
8 different climatic regions and pollution levels. Data over these four study areas were collected for the
9 years 2008 to 2017, and analyzed in a statistical sense. Both MODIS/Terra and MODIS/Aqua data were
10 used to study the difference in cloud properties between the morning and the early afternoon, i.e. with a
11 time difference of 3 hours.

12 In order to reduce differences in the initial distributions of cloud and meteorological parameters between
13 high and low AOD conditions at the start of the timestep, normalized histograms of these parameters
14 were made for high and low AOD conditions following the method described by Gryspeerd et al., (2014).
15 After that, the difference between cloud properties (CDR, COT, CWP, CF and CTP) in high and low
16 AOD conditions during the Terra overpass at 10:30 LT for each $1^\circ \times 1^\circ$ grid was investigated. We looked
17 at statistics for the 10-years dataset and found that different regions with various aerosol emission levels,
18 aerosol types and different climate characteristics show different patterns of ACI. The ACI is more
19 complex over land (BTH, YRD and PRD) than over ocean (ECS). Next, the mean change in cloud
20 properties during the 3 hours between the observations in low and high AOD conditions, as provided by
21 the differences in the observations by MODIS/Terra (morning) and MODIS/Aqua (afternoon)
22 overpasses, were examined and differences were analyzed. The results show that the COT and CWP
23 over land and ocean were increased after the 3 hours timestep, irrespective of the initial AOD
24 conditions. Furthermore, we investigated the difference between the mean change in cloud properties
25 (CDR, COT, CWP, CF and CTP) in low and high AOD conditions between the two observations. We
26 found that the variation in $d(\text{Cloud}_X)$ is different for continental and oceanic clouds. This applies to
27 CDR, cloud fraction and CTP, but not to COT and CWP. Both COT and CWP increase over land and
28 ocean after the timestep, irrespective of the AOD. The variation trend of COT and CWP after 3 hours
29 depends little on the initial AOD, but the initial AOD conditions can affect the amplitude of variation of

1 COT and CWP. Constrained by relative humidity and boundary thermodynamic and dynamic conditions,
2 the variation of $d(CF)$ in response to aerosol abundance over land was also analyzed. Two cases were
3 considered: (1) when the cloud cover increases under both low and high AOD conditions after the 3 hours
4 timestep; (2) when the cloud cover decreases under both low and high AOD conditions after the 3 hours
5 timestep. From both cases, we find that almost all $d(CF)$ values are positive, indicating that the variations
6 of CF are larger in high AOD than that in low AOD after the 3 hours timestep. The results show that
7 scenes with large cloud fraction experience large AOD and stronger upward motion of air parcels and
8 large RH when RH is larger than 20%. With regarded to the effect of LTS on the change of cloud cover,
9 scenes with large cloud fraction change experience large AOD and large LTS when LTS smaller than
10 10K. Conversely, scenes with smaller cloud fraction change experience large AOD and large LTS when
11 LTS larger than 10K and smaller than 20K. We also find that smaller cloud fraction occurs when scenes
12 experience larger AOD and larger initial cloud cover.

13 In summary, whilst we have reduced the error due to meteorological effects on aerosol retrieval,
14 meteorological covariation with the cloud and aerosol properties is harder to remove. As
15 aerosol-cloud interaction is a complex problem, it is important to synergistically use multiple
16 observation products and atmospheric models to explore the mechanisms of aerosol-cloud
17 interaction. Therefore, further analysis can be carried out in future work.

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