A water vapor modulated aerosol impact on ice crystal size

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Abstract.

The interactions between aerosols and ice clouds represent one of the largest uncertainties in global radiative forcing from pre-industrial time to the present. In particular, the impact of aerosols on ice crystal effective radius ($R_{ei}$), which is a key parameter determining ice clouds’ net radiative effect, is highly uncertain due to limited and conflicting observational evidence. Here we investigate the effects of aerosols on $R_{ei}$ under different meteorological conditions using 9-year satellite observations. We find that the responses of $R_{ei}$ to aerosol loadings are modulated by water vapor amount in conjunction with several other meteorological parameters. While there is a significant negative correlation between $R_{ei}$ and aerosol loading in moist conditions, consistent with the “Twomey effect” for liquid clouds, a strong positive correlation between the two occurs in dry conditions. Simulations based on a cloud parcel model suggest that water vapor modulates the relative importance of different ice nucleation modes, leading to the opposite aerosol impacts between moist and dry conditions. When ice clouds are decomposed into those generated from deep convection and formed in-situ, the water vapor modulation remains in effect for both ice cloud types, although the sensitivities of $R_{ei}$ to aerosols differ noticeably between them due to distinct formation mechanisms. The water vapor modulation can largely explain the difference in the responses of $R_{ei}$ to aerosol
loadings in various seasons. A proper representation of the water vapor modulation is essential for an accurate estimate of aerosol-cloud radiative forcing produced by ice clouds.

1 Introduction

Aerosols are known to interact with clouds and hence affect Earth’s radiative balance, which represents the largest uncertainty in global radiative forcing from pre-industrial time to the present (IPCC, 2013). The interactions between aerosols and liquid as well as mixed-phase clouds have been extensively studied (Rosenfeld et al., 2014; Seinfeld et al., 2016), however, much less attention has been paid to ice clouds, among which cirrus clouds are globally distributed and present at all latitudes and seasons with a global cloud cover of about 30% (Wylie et al., 1994; Wylie et al., 2005). Ice clouds, formed with various types of aerosols serving as ice nucleating particles (INPs) (Murray et al., 2012; Hoose and Moehler, 2012), act as a major modulator of global radiation budget and hence climatic parameters (e.g., temperature and precipitation) by reflecting solar radiation back to space (solar albedo effect, cooling) and by absorbing and re-emitting long-wave terrestrial radiation (greenhouse effect, warming); the balance between the two is dependent on ice cloud properties, particularly ice crystal size (Liou, 2005; Waliser et al., 2009; Fu and Liou, 1993). Limited estimates (IPCC, 2013; Liu et al., 2009; Fan et al., 2016) have shown that the global aerosol-cloud radiative forcing produced by ice clouds can be very significant but highly uncertain, ranging from –0.67 W m\(^{-2}\) to 0.70 W m\(^{-2}\). For reference purposes, the best estimate of global aerosol-cloud radiative forcing produced by all cloud types is –0.45 W m\(^{-2}\) (90% confidence interval [–1.2, 0 W/m\(^{2}\)]) according to the Intergovernmental Panel on Climate Change (IPCC) (Fig. TS.6 in IPCC, 2013).

The substantial uncertainty in aerosol-ice cloud radiative forcing arises largely from a poor understanding of the aerosol effects on ice cloud properties, in particular ice crystal effective radius (\(R_{ei}\)), a key parameter determining ice clouds’ net radiative effect (Fu and Liou, 1993). Very limited observational studies (Jiang et al., 2008; Jiang et al., 2011; Su et al., 2011; Chylek et al., 2006; Massie et al., 2007) have investigated the response of \(R_{ei}\) to aerosol loadings. Most of them (Jiang et al., 2008; Jiang et al., 2011; Su et al., 2011) found that polluted clouds involved smaller \(R_{ei}\) than clean clouds, in agreement with the classical “Twomey effect” for liquid clouds (Twomey, 1977), which states that more aerosols can result in more and smaller cloud droplets and hence larger cloud albedo. In contrast, a couple of studies over the Indian Ocean (Chylek et al., 2006; Massie et al., 2007) reported that \(R_{ei}\) is
roughly unchanged (Massie et al., 2007) or larger (Chylek et al., 2006) during more polluted episodes. It has been shown that increased aerosols (and thus INPs) lead to enhanced heterogeneous nucleation, which is associated with larger and fewer ice crystals as compared to the homogeneous nucleation counterpart (DeMott et al., 2010; Chylek et al., 2006). However, the reasons for disagreement among various studies, and the controlling factors for different aerosol indirect effects are yet to be explored, therefore the sign and magnitude of the overall aerosol effects remain in question.

With the objective to resolve the substantial uncertainty, we systematically investigate the effects of aerosols on $R_{ei}$ of two types of ice clouds under different meteorological conditions using 9-year continuous satellite observations from 2007 to 2015. The study region is East Asia and its surrounding areas (15°-55° N, 70°-135° E; Fig. S1), where aerosol loadings can range from small to extremely large values in different locations and time periods (Wang et al., 2017).

2 Data and Methods

2.1 Sources of observational data

We obtain collocated aerosol/cloud measurements primarily from MODIS (Moderate Resolution Imaging Spectroradiometer) onboard the Aqua satellite, and CALIPSO (Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations), as summarized in Table S1.

We acquire aerosol optical depth (AOD) retrievals at 550 nm from the level 2 MODIS aerosol product (MYD04, Collection 6) at a resolution of 10 km × 10 km. The accuracy of AOD (denoted by $\tau$) retrievals has been estimated to be about $\pm(0.05 + 0.15\tau)$ over land and $\pm(0.03 + 0.05\tau)$ over ocean (Levy et al., 2010; Remer et al., 2005). Similarly, we obtain cloud effective radius (equivalent to $R_{ei}$ in the case of ice phase) and cloud phase determined by the “cloud optical property” algorithm from the level 2 MODIS cloud product (MYD06, Collection 6) at a 1 km × 1 km resolution (Platnick et al., 2015).

The CALIPSO satellite flies behind Aqua by about 75 seconds and carries CALIOP (Cloud-Aerosol Lidar with Orthogonal Polarization), a dual-wavelength near-nadir polarization lidar (Winker et al., 2007). CALIOP has the capability to determine the global vertical distribution of aerosols and clouds. In this study, we make use of the CALIPSO level 2 merged aerosol and cloud layer product (05kmMLay, version 4.10) with an along-track resolution of 5 km and a high vertical resolution of 30-60 m below 20.2 km. The variables we employ for the investigation include aerosol/cloud layer numbers, layer base temperature,
layer top/base height, layer aerosol/cloud optical depth, feature classification flags (containing the flags of “cloud type” and “aerosol type”), and two quality control (QC) flags named the cloud aerosol discrimination (CAD) score, and extinction QC (Atmospheric Science Data Center, 2012).

To examine the impact of meteorological conditions on aerosol-$R_{ei}$ relations, we also obtain vertically-resolved pressure, relative humidity (RH), and temperature from the CALIPSO aerosol profile product (05kmAPro, version 4.10), and middle cloud layer temperature ($T_{mid}$) from the CALIPSO 05kmMLay product (version 4.10). The other meteorological parameters (see Table S1) are collected from the NCEP’s Final Analysis reanalysis data (ds083.2), which are produced at a $1^\circ \times 1^\circ$ resolution every six hours. Since Aqua and CALIPSO satellites overpass the study areas between 5:00-8:00 UTC, the ds083.2 datasets at 6:00 UTC are utilized.

2.2 Processing of observational data

In the analysis, we identify a CALIPSO profile layer at 5 km resolution as ice cloud when its “cloud type” is “cirrus” or its layer base temperature is colder than $-35 \, ^\circ C$. Previous studies (Mace et al., 2001; Mace et al., 2006; Kramer et al., 2016) have distinguished two major types of ice clouds characterized by distinct formation mechanisms: ice clouds generated from deep convection (convective ice clouds) and those generated in-situ due to updraft caused by frontal systems, gravity waves, or orographic waves (in-situ ice clouds). Considering that the impact of aerosols could be discrepant in different formation processes, we separate these two ice cloud types using CALIPSO data and a similar approach to that developed by Riihimaki and McFarlane (2010). First, we group ice cloud profiles at 5 km resolution into objects using the criteria that neighboring ice cloud profiles must vertically overlap (the base of the higher cloud layer is lower than the top of the lower cloud layer) and be separated by no more than 1 profiles horizontally (i.e., distance $\leq 5$ km). We subsequently classify ice cloud objects into three types, i.e., convective, in-situ, and other ice clouds, according to their connection to other clouds. The criteria to determine whether two clouds are connected are consistent with that used to group ice cloud objects, i.e., the neighboring profiles must vertically overlap and horizontally separated by no more than 5 km. Convective ice clouds consist of ice cloud objects that are connected to larger clouds that include deep convective profiles (i.e., the “cloud type” flag is “deep convection”). An ice cloud object is classified as in-situ if at least 95% of a cloud consists of a single ice cloud object which is at least 25 km (i.e., 5 profiles) in
the horizontal direction, and none of the remaining profiles are deep convection type. The remaining ice cloud objects are categorized as the “other” type. The convective, in-situ, and other ice clouds account for 44.9%, 52.4%, and 2.7% of all ice cloud profiles, respectively. The “other” type is neglected in the subsequent analysis due to very small occurrence frequency. We should be cautious that the convective and in-situ ice clouds may not be perfectly separated using the approach described above. For example, the in-situ ice clouds identified here could include convectively-detained objects that are no longer connected with their parent deep convection, and convectively-detained objects whose parent deep convective clouds do not overlap with CALIPSO’s track. The convective ice clouds may also be contaminated by some in-situ formed ice cloud objects that happen to be spatially connected to deep convection. However, the classification scheme appears to be reasonable, as indicated by the distinct properties of the two ice cloud types shown in Section 3.2. Only single-layer ice cloud profiles with valid QA flags (20 ≤ CAD score ≤ 100, Extinction QC = 0/1) are used in statistical analysis.

We then match collocated MODIS/Aqua and CALIPSO observations by averaging retrieved AOD and $R_{ei}$ from MODIS level 2 products (MYD04 and MYD06) within 30 km and 5 km radii of each 5 km ice cloud profile from CALIPSO, respectively. The averaging is done to achieve near-simultaneous aerosol and cloud measurements, since AOD observations from MODIS are missing at cloudy conditions. As AOD variation has a large spatial length scale of 40-400 km (Anderson et al., 2003), it is averaged within a larger radius than that for $R_{ei}$ to increase the number of data points with valid AOD observations. The average $R_{ei}$ is calculated based on the pixels with “cloud phase” of ice. Apart from the column AOD, we also need to obtain AOD of the aerosol layers mixed with ice cloud layers, as in-situ ice clouds are primarily affected by aerosols at the ice cloud height. For this purpose, we use the CALIPSO 05kmMLay product to select the aerosol layers which have valid QA flags (-100 ≤ CAD score ≤ -20, Extinction QC = 0/1; Huang et al., 2013) and are vertically less than 0.25 km away from the ice cloud layer following Costantino and Breon (2010). The AOD of these aerosol layers are averaged within a 30 km radius of ice cloud profiles. The meteorological parameters from the NCEP datasets (ds083.2) are matched to the CALIPSO resolution by determining which NCEP’s grid contains a certain CALIPSO 5 km profile. Finally, we eliminate profiles with column AOD > 1.5 to reduce the potential effect of cloud contamination (Wang et al., 2015).
Convective ice clouds are generated by convective updraft originating from lower troposphere and are therefore affected by aerosols at various altitudes, whereas in-situ ice clouds are primarily dependent on aerosols near the cloud height. For this reason, we use column AOD and layer AOD mixed with ice clouds as proxies for aerosols interacting with convective and in-situ ice clouds, respectively. We also investigate the overall effect of aerosols on all types of ice clouds. In this case, column AOD is used as a proxy for aerosol loading affecting ice clouds following a number of previous studies (Jiang et al., 2011; Massie et al., 2007; Ou et al., 2009). The rationale is that the MODIS-detected AOD generally shows a close correlation to the MLS (Microwave Limb Sounder)-observed CO concentration in ice clouds (Jiang et al., 2008; Jiang et al., 2009), which in turn correlates well with the aerosol loading mixed with clouds in accordance with both aircraft measurements and atmospheric modeling (Jiang et al., 2009; Li et al., 2005; Clarke and Kapustin, 2010). After the preceding screening, about $2.73 \times 10^4$, $1.09 \times 10^4$, and $5.68 \times 10^4$ profiles are used to analyze the relationships between column/layer AOD and $R_{ei}$ of convective, in-situ, and all types of ice clouds. The available profiles for in-situ ice clouds are fewer because aerosols mixed with ice clouds are often optically thin or masked by clouds and hence may not be fully detected by CALIPSO.

### 2.3 Cloud parcel model simulation

To support the key findings (i.e., the water vapor modulation of $R_{ei}$-aerosol relations) from satellite observations and elucidate the underlying physical mechanisms, we perform model simulations using a cloud parcel model, which was originally developed by Shi and Liu (2016) and updated in this study to incorporate immersion nucleation. The model mimics formation and evolution of in-situ ice clouds in an adiabatically rising air parcel. The formation of convective ice clouds involves additional physical processes, which call for more sophisticated models and future investigations. Nevertheless, as will be discussed in Section 3.4, the key processes controlling the patterns of $R_{ei}$-aerosol relations should be similar for these two ice cloud types. The model’s governing equations that describe the evolution of temperature, pressure, and mass mixing ratio, number concentration, and size of ice crystals can be found in Pruppacher and Klett (1997). The main microphysical processes considered include homogeneous nucleation and two modes of heterogeneous nucleation (deposition and immersion nucleation), depositional growth, sublimation, and sedimentation. The rate of homogeneous nucleation of supercooled sulfate droplets is calculated based on the water activity of sulfate solution (Shi and Liu, 2016). The dry sulfate aerosol is assumed to follow a
lognormal size distribution with a geometric mean radius of 0.02 μm. The deposition nucleation on externally mixed dust (deposition INP) and immersion nucleation of coated dust (immersion INP) are parameterized following Kuebbeler et al. (2014). Anthropogenic INPs are not included in the cloud parcel model following recent studies (Shi and Liu, 2016; Kuebbeler et al., 2014). This is because 1) ice nucleation experiments for black carbon show contradicting results (Hoose and Moehler, 2012), and 2) ice nucleation parameterizations for anthropogenic aerosol constituents other than black carbon have not been adequately developed under ice cloud conditions due to limited experimental data. Also, we find that the relationships between $R_{ei}$ and loadings of dust aerosols are similar to those between $R_{ei}$ and loadings of all aerosols (Section 3.1). As such, we argue that the general pattern of simulation results would remain unchanged if more INPs were incorporated. The accommodation coefficient of water vapor deposition on ice crystals is assumed to be 0.1 (Shi and Liu, 2016). The sedimentation velocity of ice crystals is parameterized following Ikawa and Saito (1991). The model neglects some ice microphysical processes such as aggregational growth of ice crystals. Although aggregational growth can affect the concentration and size of ice crystals, its effects should be relatively small in terms of the response of $R_{ei}$ to aerosol loading since this process is not strongly dependent on aerosols.

We conduct three groups of numerical experiments with discrepant available water amount for ice formation, denoted by initial water vapor mass mixing ratios (pv). Each group is comprised of 100 experiments with initial sulfate number concentrations increasing linearly from 5 cm$^{-3}$ to 500 cm$^{-3}$. For all experiments, the initial number concentrations of externally mixed dust and coated dust are prescribed to be 0.015% and 0.005% of sulfate, respectively, since INPs represent only 1 in $10^3$ to $10^6$ of ambient particles (Fan et al., 2016). The initial pressure and temperature for all experiments are set at 250 hPa and 220 K, respectively. The updraft velocity is prescribed to be 0.5 m s$^{-1}$ along with the simulation period which covers a length of 30 min. The effective radius, number concentration, and mass mixing ratio of ice crystals at the end of the experiments are used to construct the aerosol-cloud relationships.

### 3 Results and Discussion

#### 3.1 Relationships between $R_{ei}$ and aerosols modulated by meteorology

In this section we discuss the impact of aerosols on $R_{ei}$, with both ice cloud types lumped together, based on satellite data (Fig. 1). The aerosol effects on individual ice cloud types will be discussed in the next section. The dash line in Fig. 1a shows the overall changes in $R_{ei}$ with
AOD. R_{ei} generally increases with increasing AOD for moderate AOD range (< 0.5), and decreases slightly for higher AOD. This relationship is attributable to complex interactions between meteorological conditions and microphysical processes, which will be detailed below.

Having shown overall response of R_{ei} to AOD, we investigate whether the responses are similar under different meteorological conditions. We plot the R_{ei}-AOD relationships separately for different ranges of meteorological parameters, as shown in Fig. 1a-c and Fig. S2. Included in the analysis are most meteorological parameters that can potentially affect ice cloud formation and evolution, including the RH averaged between 100 hPa and 440 hPa, convective available potential energy (CAPE) which is an indicator of convective strength, middle cloud layer temperature (T_{mid}), wind speed and direction at ice cloud height and at surface, vertical velocity below and at ice cloud height, and vertical wind shear. For some meteorological parameters, e.g., vertical wind shear and vertical velocity at 300/500 hPa, the curve shapes are similar for different meteorological ranges. However, for RH, CAPE, and U-component of wind speed at 200 hPa (U200), the curve shapes vary significantly according to different ranges (Fig. 1a-c). As illustrated by RH and CAPE, R_{ei} decreases significantly with increasing AOD for high RH (> 65%) or CAPE (> 500 J/kg) following the rule of “Twomey effect”. In contrast, for low RH (< 45%) or CAPE (0 J/kg), R_{ei} increases sharply with AOD.

To the best of our knowledge, the strong dependency of R_{ei}-AOD relationships on meteorological conditions for ice clouds has been demonstrated for the first time.

These correlations, however, may not be necessarily attributed to aerosols. It is theoretically possible that certain meteorological parameters lead to simultaneous changes in both AOD and ice cloud properties and produce a correlation between these two parameters. To rule out this possibility, we examine the responses of AOD to the above-mentioned meteorological parameters (Fig. S3) and find that AOD does not serve as proxy for them since it varies by less than 0.2 in response to variation in any meteorological parameter. Furthermore, we bin observed R_{ei} according to RH, CAPE, and U200, for different ranges of AOD (Fig. 1d-f). Using RH as an example, a larger AOD corresponds to smaller R_{ei} for a given RH within the larger RH range, whereas an increase in AOD enlarges R_{ei} for a given RH within the smaller RH range. Similar results are found for CAPE and U200 (Fig. 1d-f), demonstrating the role of aerosols in altering R_{ei} under the same meteorological conditions.

Moreover, the cloud contamination in AOD retrieval (Kaufman et al., 2005) or aerosol contamination in cloud retrieval (Brennan et al., 2005) is not likely to lead to observed R_{ei}-AOD correlations, because the retrieval biases cannot explain the opposite correlations under...
different meteorological conditions. Therefore, we conclude that both the positive and negative correlations between AOD and $R_{ei}$ are primarily attributed to the aerosol effect. This causality is also supported by numerical simulations using a cloud parcel model to be described in Section 3.4. Furthermore, we find that the three meteorological parameters which pose the strongest impact on $R_{ei}$-AOD relationships (RH, CAPE, and U200) are closely correlated with each other, with correlation coefficients between each two exceeding ±0.5 and p-value less than 0.01 (Table S2). In fact, all these three parameters are closely related to the amount of water vapor available for ice cloud formation. It is obvious that RH averaged between 100-440 hPa is an indicator of water vapor amount. CAPE represents convective strength and hence water vapor lifted to ice cloud heights; U200 denotes the origin of air mass such as moist Pacific Ocean (easterly wind) or dry inland continent (westerly wind). Therefore, water vapor amount is likely a key factor which modulates the observed impact of aerosols on $R_{ei}$.

The proposed mechanism for the water vapor modulation is that different water vapor amount substantially alters the relative significance of different ice nucleation modes, thereby resulting in discrepant $R_{ei}$-AOD relationships. Specifically, ice crystals form via two primary pathways: homogeneous nucleation of liquid water droplets (or supercooled solution particles) below about −35 °C, and heterogeneous nucleation triggered by INPs (IPCC, 2013; DeMott et al., 2010). INPs possess surface properties favorable to lowering the ice supersaturation ratio required for freezing (IPCC, 2013; DeMott et al., 2010), therefore the onset of heterogeneous nucleation is generally easier and earlier in rising air parcels. Under moist conditions, homogeneous nucleation dominates and thus more aerosols could lead to the formation of more and smaller ice crystals, which is in connection with the “Twomey effect” for liquid clouds. Under dry conditions, however, the earlier onset of heterogeneous nucleation can strongly compete with and possibly prevent homogeneous nucleation involving more abundant water-bearing or solution particles (IPCC, 2013; DeMott et al., 2010). Therefore, more aerosols (and hence more INPs) are expected to lead to a higher fraction of ice crystals produced by heterogeneous nucleation comprising of fewer and larger ice crystals (“anti-Twomey effect”). These proposed mechanisms will be supported and elaborated on using model simulations in Section 3.4.

Here an inherent assumption is that INP concentration is roughly proportional to, or at least positively correlated with AOD. Considering that INPs only account for a small fraction of ambient aerosols, we may not take this assumption for granted. Here we plot the $R_{ei}$-AOD...
relations using only the cases in which the “aerosol type” (a flag contained in the feature
classification flags of CALIPSO) is dust (Fig. 1g-i), and find that the water modulation effect
is very similar to the preceding results (i.e., Fig. 1a-c). Since specific components of dust
aerosols have been known as effective INPs (Murray et al., 2012; Hoose and Moehler, 2012),
the similar $R_{ei}$-AOD relations of dust and of all aerosols to some extent support the proposed
mechanisms for water vapor modulation.

3.2 $R_{ei}$-aerosol relationships for two types of ice clouds

Considering that distinct formation mechanisms of convective and in-situ ice clouds may lead
to different aerosol effects, we distinguish these two ice cloud types based on their connection
to deep convection (Section 2.2). Figure 2 illustrates the accumulative probability distribution
of cloud thickness, cloud optical thickness (COT), and $R_{ei}$ of the two ice cloud types. The
cloud thickness and COT of convective ice clouds are remarkably larger than those of in-situ
ice clouds, because more water is transported to upper troposphere in the formation process of
the former type, consistent with numerous aircraft measurement results (e.g., Kramer et al.,
2016; Luebke et al., 2016; Muhlbauer et al., 2014). The $R_{ei}$ of convective ice clouds is slightly
larger than that of in-situ ice clouds, which has also been reported in a number of aircraft
campaigns (Luebke et al., 2016; Kramer et al., 2016). The larger $R_{ei}$ in convective ice clouds
is attributed to larger water amount and the fact that they are produced by convection
emerging from lower altitude. Below the $-35 \, ^\circ C$ isotherm, ice crystals stem only from
heterogeneous nucleation, which tends to produce larger ice crystals compared to the
homogeneous nucleation counterpart (Luebke et al., 2016).

Figures 3 and 4 show the impact of aerosols on $R_{ei}$ under different meteorological
conditions for convective and in-situ ice clouds, respectively. As described in Section 2.2, we
use column AOD and layer AOD mixed with ice clouds as proxies of aerosols interacting
with convective and in-situ ice clouds, respectively. The most impressive feature from these
figures is that the meteorology modulation remains in effect for either of the two ice cloud
types, such that $R_{ei}$ generally decreases with AOD under high RH/high CAPE/negative $U_{200}$
conditions, whereas the reverse is true under low RH/low CAPE/positive $U_{200}$ conditions.
Similar to the Section 3.1, we also demonstrate that the $R_{ei}$-aerosol relationships are primarily
attributed to the aerosol effect by illustrating role of aerosols in altering $R_{ei}$ under the nearly
constant meteorological conditions (Fig. 3d-f, Fig. 4d-f). For example, a larger AOD is
associated with a smaller $R_{ei}$ for a given RH within the larger RH range, while an increase in
AOD leads to a larger $R_{ei}$ for a given RH within the smaller RH range. These results illustrate
that the meteorology modulation of aerosol effects on $R_{ei}$ is valid regardless of ice cloud formation mechanisms.

A closer look at Figs. 3 and 4 shows that there exist noted differences between the $R_{ei}$-aerosol relationships for the two ice cloud types. For convective ice clouds, a strong positive correlation between $R_{ei}$ and AOD is found under dry conditions, while a weaker negative correlation is found under moist conditions. For in-situ ice clouds, however, weaker positive and stronger negative correlations are shown under dry and moist conditions, respectively. As a result, overall $R_{ei}$ slightly increases with increasing aerosol loading for convective ice clouds, but slightly decreases for in-situ clouds. These differences are again linked to the distinct formation mechanisms of the two ice cloud types. Convective ice clouds are influenced by aerosols at various heights, which presumably contain much more INPs than the thin upper tropospheric aerosol layers in the case of in-situ ice clouds. In addition, the heterogeneously formed ice crystals in convective clouds are able to grow before being lifted to the $-35 \, ^\circ C$ isotherm, giving rise to a larger difference between the ice crystal sizes produced by heterogeneous and homogeneous nucleation as compared to in-situ ice clouds.

For these reasons, under dry conditions, the increase in $R_{ei}$ with aerosol loading, which is due to the transition from homogeneous-dominated to heterogeneous-dominated regimes, would be much more pronounced for convective ice clouds. At moist conditions, although homogeneous nucleation dominates for both ice cloud types (see Section 3.1), the mass fraction of homogeneously formed ice crystals is smaller for convective ice clouds, leading to a weaker decline in $R_{ei}$ with aerosols.

### 3.3 Seasonal variations in $R_{ei}$-aerosol relationships

Furthermore, we find that the meteorological modulation can largely explain differences in $R_{ei}$-AOD relationships as a function of season. Figure 5a shows that the $R_{ei}$-AOD relationships are dramatically different associated with season, such that $R_{ei}$ decreases significantly with increasing AOD in summer (June, July, and August), while $R_{ei}$ increases rapidly in winter (December, January, and February). Figure 5d-f illustrate the probability distribution functions (PDFs) of RH, CAPE, and U200 in different seasons (the area under any PDF equals 1.0). The overlapping area of PDFs in summer and winter represents the degree of difference in meteorological conditions between these two seasons. We find that meteorological conditions are significantly distinct in summer and winter in terms of RH, CAPE, and U200, as indicated by relatively small overlapping areas (<0.6) for these three parameters. The RH and CAPE tend to be higher and U200 tends to be more negative in
summer. Moreover, the shapes of $R_{ei}$-AOD curves in summer and winter highly resemble those under high-RH/high-CAPE/negative-U200 and low-RH/low-CAPE/positive-U200 conditions, respectively (see Fig. 1a-c), which demonstrates that the discrepancy in meteorological conditions between winter and summer can, to a large extent, explain the distinct $R_{ei}$-AOD relationships in these two seasons. Furthermore, when ice clouds are decomposed into convective and in-situ types, Fig. 5b-c show that the above-mentioned patterns hold true for both types, indicating that the seasonal variations in $R_{ei}$-aerosol relations for both ice cloud types are largely attributable to the meteorology modulation.

### 3.4 Modeling support for the water vapor modulation

We have shown that the $R_{ei}$-aerosol relationships are modulated by meteorological conditions, particularly water vapor amount. To support the observed relationships and our proposed physical mechanisms, we perform model simulations as described in Section 2.3 and summarize the results in Fig. 6.

Figure 6a reveals that the simulated patterns of $R_{ei}$-aerosol relationships under different water vapor amount agree well with the corresponding observed patterns (Fig. 1a-c). Specifically, with adequate water vapor ($pv = 106$ ppm), $R_{ei}$ decreases significantly with aerosol concentrations (“Twomey effect”). Under a dry condition ($pv = 38$ ppm), $R_{ei}$ increases noticeably with aerosol concentrations (“anti-Twomey effect”). With a medium water vapor amount ($pv = 53$ ppm), $R_{ei}$ first decreases and then increases. A deeper analysis of the simulation results supports our proposed mechanism (Section 3.1) that the competition between different ice nucleation modes is the key to explain the water vapor modulation. With adequate water vapor ($pv = 106$ ppm), the onset of deposition and immersion nucleation under relatively lower ice supersaturation consumes only a small fraction of water vapor due to the small INP population, and homogeneous nucleation acts as the dominant ice formation pathway, as indicated by the very small number/mass fraction (< 1%) of heterogeneously formed ice crystals shown in Fig. 6b. In this case, more aerosols are associated with the formation of more and smaller more and smaller ice crystals, consistent with the simulation results of Liu and Penner (2005). With inadequate water vapor ($pv = 38$ ppm), Fig. 6b reveals that the number fraction of heterogeneously formed ice crystals increases dramatically from about 30% to nearly 100% when aerosol number concentrations increase from 5 cm$^{-3}$ to ~300 cm$^{-3}$ (the INP number concentrations increase proportionally). This indicates that the water vapor consumed by heterogeneous nucleation is large enough to suppress or even prevent homogeneous nucleation that occurs spontaneously over a higher and narrow ice
supersaturation range (140-160%). Since the outburst of homogeneous nucleation generally
produces more ice crystals at smaller size compared with the heterogeneous counterpart, an
increasing fraction of heterogeneous nucleation would result in fewer ice crystals with larger
average size.

The current cloud parcel model simulates the environmental conditions and physical
processes for in-situ ice clouds. Although the formation processes of convective ice clouds are
considerably different from those of in-situ ones, the competition between homogeneous and
heterogeneous nucleation for water vapor hold true for both ice cloud types, and is very likely
to explain the water vapor modulation of $R_{ei}$-aerosol relations for both types. Nevertheless,
more simulation studies are still needed to support and better understand the discrepant
sensitivities of $R_{ei}$ to aerosols for these two cloud types.

As a simplified model, the simulation results of the cloud parcel model may not be
directly compared with the satellite observations. For example, we use column/layer AOD and
RH averaged between 100-440 hPa (and CAPE, U200) as proxies for aerosol loading and
available water amount, respectively, in satellite data analysis. In contrast, the concentrations
of aerosols and water vapor within a single air parcel are employed in the model simulation.
A direct comparison between satellite observations and model simulations would require
running a 3-D atmospheric model, which calls for further in-depth studies. It is also noted that
the model-simulated magnitude of $R_{ei}$ is generally smaller than observations probably due to
limited simulation periods, exclusion of ice aggregational growth and water vapor uptake
from outside of the air parcel, and variability in updraft velocity (Kramer et al., 2016;
Feingold, 2003); nevertheless, the model has successfully reproduced the key microphysical
processes and hence captured the observed patterns of $R_{ei}$-aerosol relationships.

4 Conclusions and implications

In this study, we investigate the effects of aerosols on $R_{ei}$ under different meteorological
conditions using 9-year satellite observations. We find that the responses of $R_{ei}$ to aerosol
loadings are modulated by water vapor amount in conjunction with several other
meteorological parameters, and vary from a significant negative correlation (“Twomey effect”)
to a strong positive correlation (“anti-Twomey effect”). Simulations using a cloud parcel
model indicate that the water vapor modulation works primarily by altering the relative
importance of different ice nucleation modes. The water vapor modulation holds true for both
convective and in-situ ice clouds, though the sensitivities of $R_{ei}$ to aerosols differ noticeably
between these two ice cloud types due to distinct formation mechanisms. The water vapor
modulation can largely explain the different responses of $R_{ei}$ to aerosol loadings in various seasons.

$R_{ei}$ is a key parameter determining the relative significance of the solar albedo (cooling) effect and the infrared greenhouse (warming) effect of ice clouds; the variation of $R_{ei}$ could change the sign of ice clouds’ net radiative effect (Fu and Liou, 1993). Aerosols have strong and intricate effects on $R_{ei}$ through their indirect effect. We provide the first and direct evidence that the competition between the “Twomey effect” and “anti-Twomey effect” is controlled by certain meteorological parameters, primarily water vapor amount. Consequently, the first aerosol indirect forcing, defined as the radiative forcing due to aerosol-induced changes in $R_{ei}$ under a constant ice water content (IPCC, 2013; Penner et al., 2011), would change from positive to negative between high and low RH ranges, implying that the water vapor modulation could play an important role in determining the sign, magnitude, and probably seasonal and regional variations of aerosol-ice cloud radiative forcings. An adequate and accurate representation of this modulation in climate models will undoubtedly induce changes in the magnitude and sign of the current estimate of aerosol-ice cloud radiative forcing. Finally, although this study focuses on East Asia, we anticipate that the present findings might be generalized to other regions as well in view of the fact that the aerosol loadings in East Asia usually span a larger range than other regions and that the aerosol effects on ice cloud properties are particularly pronounced at low and moderate aerosol loadings (Figs. 1, 3, 4, 5).

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References


Figure 1. Influence of meteorological conditions on aerosol induced variability in ice crystal effective radius ($R_{ei}$) of ice clouds. (a-c) Changes in $R_{ei}$ with AOD for different ranges of (a) RH averaged between 100 hPa and 440 hPa, (b) CAPE, and (c) U200. (d-f) Changes in $R_{ei}$ with (d) RH, (e) CAPE, and (f) U200 for different ranges of AOD. (g-i) The same as (a-c) but for the profiles with dust aerosols only. The meteorological parameters and AOD are divided into 3 and 2 ranges containing similar numbers of data points, respectively; the curves for the medium meteorological range are not shown. The error bars denote the standard errors ($\sigma/\sqrt{N}$) of the bin average, where $\sigma$ is the standard deviation and N is the sample number. The influences of other meteorological parameters are shown in Fig. S2.
Figure 2. Accumulative probability distribution of the properties of two ice cloud types: (a) cloud thickness, (b) cloud optical thickness, and (c) $R_{ei}$.

Figure 3. The same as Fig. 1a-f but for convective ice clouds.
Figure 4. The same as Fig. 1a-f but for in-situ ice clouds. Note that we use AOD of the aerosol layers mixed with ice clouds rather than column AOD, since in-situ ice clouds are primarily affected by aerosols at the ice cloud height.
Figure 5. Changes in $R_{ei}$ with AOD and the probability distribution of selected meteorological parameters as a function of season. (a-c) Changes in $R_{ei}$ with AOD as a function of season for (a) all ice clouds, (b) convective ice clouds, and (c) in-situ ice clouds. (d-f) The probability distribution of (d) RH averaged between 100 hPa and 440 hPa, (e) CAPE, and (f) U200 as a function of season. Definitions of season are as follows: Winter – December, January, and February; Spring – March, April, and May; Summer – June, July, and August; Fall – September, October, and November. The definition of error bars is the same as in Fig. 1.
Figure 6. Simulated changes in (a) $R_{ei}$ and (b) fraction of ice crystals produced by the heterogeneous nucleation as a function of the total aerosol number concentration. Simulations are conducted for three initial water vapor mass mixing ratios ($pv$), an indicator of available water amount for ice formation. The ratios of externally mixed dust (deposition INP), coated dust (immersion INP), and sulfate (not INP) are prescribed with values of 1.5:0.5:10000 in all experiments.