THE AEROSOL INDIRECT EFFECTS EXAMINED BY NUMERICALLY CALCULATED AEROSOLS
AND SATELLITE DERIVED CLOUDS

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1. INTRODUCTION

Twomey (1977) first described the modification of cloud properties due to aerosols. Additional particles, which serve as cloud condensation nuclei (CCN), lead to a decrease in cloud particle size and an increase in the cloud optical depth of the constant liquid water content (LWC). This change is known as the aerosol indirect effect of the first kind, and the interaction between aerosols and clouds is one of the least understood processes related to climate change, including changes associated with global warming. The uncertainty of estimated global radiative forcing due to the aerosol indirect effect ranges from 0 to −2 W/m², and the confidence level is “very low” (Ramaswamy 2001). It is therefore critical to accumulate research examples and provide a more reliable quantitative estimate of radiative forcing that accounts for the aerosol indirect effect.

* A number of studies have attempted to monitor this

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For example, Han et al. (1994) used global data collected by the Advanced Very High Resolution Radiometer (AVHRR) to describe water cloud particle size. They found a distinct difference between land and ocean in cloud effective particle size. The number and chemical composition (e.g., hygroscopicity) of aerosols play important roles in determining cloud particle size. Feingold et al. (2003) used ground-based radiometric measurements and showed that the water cloud effective particle size decreases as the aerosol index increases. Nakajima et al. (2001) used AVHRR data to study the relationship between aerosols and water clouds over the global ocean. They showed that cloud effective particle sizes decrease, but the cloud optical depth increases, as the aerosol number density increases. Subsequently, Sekiguchi et al. (2003) described the behaviors of additional cloud properties (such as the cloud amount and cloud top temperature) using the aerosol number density based on the same AVHRR data set used by Nakajima et al. (2001). Breon et al. (2002) used Polarization and Directionality of the Earth’s
Reflectances (POLDER) data to analyze aerosol–cloud relationships over land and ocean. They documented different rates of change in cloud effective particle size as a function of the aerosol indices for land and ocean.

In addition to global-scale analyses such as those mentioned above, regional studies targeting Asia are important due to high emission rates and locally focused aerosols. Chameides et al. (2002) compared model-simulated aerosol loading to cloud properties derived from International Satellite Cloud Climatology Project (ISCCP) data, with a focus over China. The ISCCP data included optical depth and showed that cloud optical depth positively correlated with aerosol loading. The relationship between aerosol loading and cloud amount, however, was not clear. They did not consider the behavior of cloud microphysical parameter because of limited ISCCP products.

Industrial production in China since the 1980s has quickly expanded following substantial political changes; consequently, various gaseous emissions have also rapidly increased (Streets and Waldhoff 2000). Heightened industrial activity has generated many aerosols and influenced air quality. China is therefore an ideal site to study the relationships between anthropogenic aerosols and clouds. As in the investigation by Chameides et al. (2002), this study examined the relationships between anthropogenic aerosols and clouds, including microphysics, to better understand human-induced effects on clouds over East Asia. Similar to the above-referenced papers (e.g., Nakajima et al. 2001, Breon et al. 2002, Sekiguchi et al. 2003), the present work also took a correlative approach to examine how aerosols affect the behaviors of cloud properties, rather than pursuing the causative mechanisms of the aerosol indirect effect. Data used in this work included numerically simulated anthropogenic aerosol concentration and satellite-derived cloud properties. Both data types are described in the next section. A discussion of relationships between aerosols and clouds and the mechanisms involved follows. The last section provides the conclusions.

2. DATA

2.1. Anthropogenic aerosol concentration calculated from numerical model simulations

Satellite retrieval of aerosol properties has been considered difficult due to high reflectivity in visible wavelengths over land surfaces (Husar et al. 1997) and by cloud contamination in pixels. Therefore, this study used results from numerical simulations of anthropogenic aerosol concentration to examine the effects of human activities. Numerical simulations can provide long-term archives without gaps in data and can extract only anthropogenic components. The three-dimensional regional-scale chemical transport model used in this study was based on the Models-3 Community Multiscale Air Quality (CMAQ) modeling system released by the United States Environmental Protection Agency (Byun and Ching 1999) targeting East Asia (15–55°N, 90–135°E) with an 80 x 80-km² grid resolution. For vertical resolution, there were 14 layers in the z-coordinate system up to 23 km, with approximately seven layers within the boundary layer below 2 km. The meteorological fields were obtained.
from the Regional Atmospheric Modeling System (RAMS) version 4.3 (Pielke et al. 1992; Cotton et al. 2003) using The National Center for Environmental Protection/National Center for Atmospheric Research (NCEP/NCAR) 2.5 x 2.5° reanalysis data sets for every 6 h; validation of this simulation was confirmed by Uno et al. (2005). The Statewide Air Pollution Research Center, version 99 (SAPRC-99) chemical mechanism, which includes 72 species and 214 reactions (Carter 2000) with an Euler Backward Iterative (EBI) solver, was adopted for gas phase chemistries in the CMAQ system used in this study. This CMAQ/RAMS Asia-scale modeling system was developed by Sugata et al. (2001) and Uno et al. (2005). Initial conditions for the chemical species modeled were based on the lowest values of the observed range. Boundary conditions for CMAQ were taken from monthly average output from a coupled global chemistry–climate model (Chemical AGCM for Study of Atmospheric Environment and Radiative Forcing: CHASER; Sudo et al. 2002a, b). For this CMAQ modeling system, anthropogenic emissions were obtained from an emission inventory for 1995 with a 0.5° x 0.5° resolution developed by the Frontier Research System for Global Change (FRCGC) and Research Institute for Humanity and Nature (RIHN) [http://www.jamstec.go.jp/forsrce/research/d4/emission.htm]. The fraction generated by biomass burning was sufficiently small so that the calculated aerosols could be considered largely “anthropogenic aerosols.” Non-methane volatile organic compounds (NMVOC) emissions were divided into the lumped-hydrocarbon SAPRC-99 categories using the method of Carter (2000). The five chemical species (sulfates, nitrates, ammonia, and elemental and organic carbon) were considered to define the vertically integrated anthropogenic aerosol mass concentration $M_a$ ($\mu g/m^3$) in this work. Respective aerosol particles were assumed to have the following three log-normal size distributions: an Aitken mode having a geometrical mean radius of 0.03 $\mu m$ and geometrical standard deviation of 1.7, an accumulation mode having a geometrical mean radius of 0.3 $\mu m$ and geometrical standard deviation of 2.0, and a coarse mode having a geometrical mean radius of 6.0 $\mu m$ and geometrical standard deviation of 2.2.

Zhang et al. (2003) confirmed that simulated SO$_4$ concentrations agreed well with in situ observations in most cases but were overestimated in conditions such as high humidity in summer, weak rain, and fog. Simulations for 1995 were performed, and monthly mean values for January, April, July, and October were stored for this work.

2.2. Low cloud optical and microphysical properties observed using satellite data

Satellite data used in this work defined the cloud properties. The algorithm by Kawamoto et al. (2001) was used to retrieve cloud optical depths at visible wavelength $\tau$, effective particle radius $r_e$ ($\mu m$), and cloud top temperature $T_t$ (K) warmer than 273 K. Cloud-reflected visible and near-infrared radiances and cloud-emitted infrared radiance were used. The retrieval assumed that solar radiance reflected by clouds at nonabsorbing visible wavelengths was a function of $\tau$ and that solar radiance reflected by clouds
at water-absorbing near-infrared wavelengths was a function of \( r_v \) near the cloud top (Nakajima and King 1990). The NCEP/NCAR reanalysis data provided humidity, pressure, temperature profiles, and surface temperatures to compute the cloud-emitted and surface-emitted thermal radiations and water vapor amounts above and under the cloud layer in the retrieval process.

Only low-level water clouds with cloud top temperatures warmer than 273 K may have occurred for the following reasons: most aerosols are within the lower part of the troposphere, and thus, aerosol interactions are more likely in low clouds than in other cloud types such as mid- and high-level clouds. Furthermore, this algorithm used the Mie scattering theory, which only applies to spheres in radiative transfer calculations. Super-cooled cloud particles below 273 K (until around 260 K) may also exist, but in an attempt to avoid potential bias in the correlations due to the inclusion of ice crystals, “clouds” hereafter refers to water clouds unless otherwise noted. According to Kawamoto et al. (2001), retrieval errors are generally 5–10% for \( \tau \) and 10–15% for \( r_v \), respectively.

The algorithm was applied to AVHRR Global Area Coverage (GAC) 3-channel radiances for 1995 (from NOAA-14), a year with relatively less influence from the Mt. Pinatubo eruption of 1991 and major El Niño Southern Oscillation (ENSO) events. The central wavelengths of channels 1, 3, and 4 of AVHRR were 0.64, 3.73, and 10.8 \( \mu \)m, respectively. East Asia (15°–55°N, 90°–135°E) was divided into 0.5° boxes to facilitate efficient processing. Each box contained 100 pixels. The segmented data were constructed daily for the analysis period, and the one analyzed pixel had the median value of visible reflectivities among the cloudy pixels in each box. The cloud-screening procedure used visible and infrared radiances following the method of Kawamoto et al. (2001). The data processing incorporated the results by Iwabuchi and Hayasaka (2002), who showed that shadowing and three-dimensional effects are substantially reduced at viewing angles smaller than approximately 25°. Months that are representative of each season (January, April, July, and October) were analyzed, but temperature restrictions (warmer than 273 K) and solar zenith angle restrictions (smaller than 70°) in January yielded too few retrievals to determine statistical significance. Therefore, only the April, July, and October results were considered in this study. Calibration constants for channel 1 were adopted from Rao and Chen (1996) due to degradation after launch. Those for channels 3 and 4 were taken from internal black body instruments. The minimum value of visible radiance during the month in each grid box was converted to the ground reflectivity.

3. RESULTS AND DISCUSSION

Values of \( M_r \) and satellite-derived \( \tau \) and \( r_v \) were correlated quantitatively in July as follows to examine cloud behaviors in relation to anthropogenic aerosols over East Asia. First, values of \( M_r \) were subdivided into bins. Cloud properties in the same geographical grid as the aerosol concentration were then collected, and an average and standard deviation of cloud properties for each \( M_r \) bin were computed. This comparison method has been used in previous studies (Nakajima et al.
Figures 1 and 2 present the relationships of \( \tau \) and \( r_c \), respectively, for each \( M_a \) in July, with standard deviations shown. The two figures show that \( \tau \) increased and \( r_c \) decreased as \( M_a \) increased. As mentioned previously, these tendencies are out of the range of retrieval uncertainty noted by Kawamoto et al. (2001). Such behaviors are consistent with the Twomey effect, i.e., the aerosol indirect effect of the first kind. As more aerosols are released to the atmosphere, more particles can serve as CCN and more cloud droplets are generated. In such situations, cloud droplet sizes will remain small because the constant LWC suggests that less water is available for each cloud droplet. The cloud optical depth, which is equivalent to the total surface area of cloud droplets, would increase due to the stronger effect of droplet number enhancement than that of particle size reduction. Other conditions such as aerosol hygroscopicity, dynamical processes, and water supply would also influence cloud properties in actual situations, as described in section 3.2. The \( \text{SO}_2 - r_c \) relationship reported by Kawamoto et al. (2004) in which \( \text{SO}_2 \) is converted to sulfate aerosols and then to CCN is analogous to the relationships proposed in this paper. Change rates of both \( \tau \) and \( r_c \) with \( M_a \) became less steep for higher values of \( M_a \) (more than approximately 40 \( \mu g/m^2 \)), indicating saturation of the aerosol–cloud interaction in that region. A few previous studies have taken a similar correlative approach to the relationships between aerosols and clouds. For example, Chameides et al. (2002) used modeled anthropogenic aerosols and satellite-derived \( \tau \) and found a comparable tendency to that in the present study. Chameides et al. (2002) showed an overall positive correlation starting at a \( \tau \) value of about 5 for an \( M_a \) of approximately 10 \( \mu g/m^2 \) to a \( \tau \) of approximately 12 for an \( M_a \) of about 50 \( \mu g/m^2 \), with saturation for large \( M_a \) values.

Regarding \( r_c \), Breon et al. (2002) found that aerosol indices were roughly 0.35–0.4 for heavily polluted regions such as the coastal and inland industrial regions of China and 0.08–0.12 for suburban regions. The values of \( r_c \) were approximately 7.7 \( \mu m \) and 8.3 \( \mu m \), respectively, for these regions. Geographically, such conditions corresponded to an \( r_c \) of approximately 9 \( \mu m \) for an \( M_a \) over 55 \( \mu g/m^2 \) and an \( r_c \) of approximately 11–9.3 \( \mu m \) for an \( M_a \) of 10 to 20 \( \mu g/m^2 \) in our case.

Breon et al. (2002) focused on the period from March to May using satellite-derived aerosol data that included both natural and anthropogenic origins.
Therefore, simple comparison of changing rates with our study is difficult. Differences in satellite sensors also raise problems. Rosenfeld and Feingold (2003), for example, have noted the differences in cloud sampling preferences between AVHRR and POLDER. The results of the present study are also difficult to compare with those by Nakajima et al. (2001) due to the type of satellite aerosol retrieval (including natural and anthropogenic aerosol origins) and because Nakajima et al. collected data only over the ocean. Among other satellite sensors, recent advancements in the Multiangle Imaging SpectroRadiometer (MISR) have led to unique monitoring at various angles that appear promising for land aerosol retrievals (Diner et al. 2001).

This study also investigated the relationship between $M_d$ and the vertically integrated cloud droplet number $N_c \, (1/cm^3)$, even though $N_c$, which was estimated from $\tau$ and $r_c$, was not an independent parameter. The following procedure was used to determine $N_c$. A log-normal size distribution with a variance of 0.35 was assumed for cloud droplets. The number density (1/cm$^3$) was calculated by integrating the size distribution; the number density was then divided by the extinction coefficient (1/cm) at the wavelength at which $\tau$ was observed to obtain the number density per optical depth (1/cm$^2$). Then $N_c$ was determined as the product of this quantity and $\tau$. As Fig. 3 presents, when $M_d$ increased, $N_c$ also increased. The vertically integrated aerosol particle number $N_a$, which was determined by integrating assumed size distributions with mass constants for species and $M_a$, was proportional to $M_d$; therefore, a similar positive relationship can be expected between $N_a$ and $N_c$. This result can be considered to confirm one aspect of the Twomey effect (i.e., an increase in cloud droplet number with an increase in aerosol particle number).

Cloud property differences can be investigated with respect to differences in cloud top heights. Water clouds were divided into middle clouds (cloud top heights >3 km) and lower clouds (cloud top heights <3 km), and the same procedures as above were followed for both cloud types. Figures 4 and 5 show the relationships between $\tau$ and $M_a$, and $r_c$ and $M_a$, for lower and middle clouds in July, respectively. Absolute values of both $\tau$ and $r_c$ were generally larger for middle clouds. Cloud optical depth is a vertically integrated quantity, so higher clouds have a greater chance to become thicker. Larger cloud particle sizes in middle clouds arise from the smaller aerosol numbers at that altitude (note that this remote sensing method retrieves $r_c$ near the cloud top), as indicated by Kawamoto et al. (2001). Increases in $\tau$ and decreases in $r_c$ and $M_a$ for both cloud types were similar to those in Figs. 4 and 5 and also consistent with the Twomey effect. However, geographical patterns between $M_d$ and cloud properties in April over southern China (around
22–32°N, 105–120°E) were not as comparable. This suggests that not only the particle processes considered here but also dynamic processes should be examined because this area is rather convective.

4. Conclusions

This paper has discussed the relationships between anthropogenic aerosols and low-level water clouds to try to better understand the aerosol indirect effect over East Asia. Results from numerical simulations yielded the anthropogenic aerosol concentration. Satellite-derived products yielded information on low-level water cloud properties. Comparisons of monthly means for aerosols and clouds showed that \( \tau \) increased and \( r_c \) decreased as \( M_a \) increased. Such tendencies were consistent with the Twomey effect, which describes how aerosols affect cloud properties. This behavior is analogous to the \( \text{SO}_2 \)–\( r_c \) relationship described by Kawanoto et al. (2004). Nevertheless, comparisons of \( M_a \) with \( \tau \) in April and October would suggest the importance of dynamic effects on cloud formation and maintaining processes that were not intensively discussed in this work. Analyses that combine dynamical processes will be the subject of further study.

Values of \( N_c \) that were calculated from \( \tau \) and \( r_c \) also increased as \( M_a \) increased. This result also agreed with the Twomey effect by indicating that additional aerosols generated more cloud droplets. A comparison of \( M_a \) with lower and middle clouds revealed similar tendencies to the previous case (i.e., total water cloud case), but differences in \( r_c \) (i.e., larger for middle and smaller for lower clouds) reflected the vertical profile of aerosol numbers. However, differences in \( \tau \) (i.e., thicker for middle and thinner for lower clouds) might have been influenced by the vertical extent. These findings were obtained with cloud microphysical information that was unavailable to Chameides et al. (2002).

The changing rates of cloud properties with \( M_a \) are important not only to describe the phenomena but also for parameterizations in numerical model applications. More studies involving both satellite observations and numerical modeling are needed to validate these rates and make them more robust. There are some papers that support the conclusions of this study. Breon et al. (2002), for example, stated that “there is a clear physical process that predicts such a decrease [of cloud particle size]…this observed correlation can be
interpreted as a causal connection.” Although we showed several phenomena that were consistent with the effect suggested by Twomey, the correlative analysis that we used here does not guarantee a causal relationship between the two quantities. Therefore, we must exercise caution in discussing the robust causality of the aerosol indirect effect and must continue to accumulate research examples based on observations and modeling.

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**Figure captions**

Fig. 1. Relationship between $M_a$ and $t$ in July 1995. 

Fig. 2. Relationship between $M_a$ and $r$ in July 1995. 

Fig. 3. Relationship between $M_a$ and $N$ in July 1995. 

Fig. 4. Relationship between $M_a$ and $\tau$ for middle and lower clouds in July 1995. 

Fig. 5. Relationship between $M_a$ and $r$ for middle and lower clouds in July 1995.