An investigation into seasonal and regional aerosol characteristics in East Asia using model-predicted and remotely-sensed aerosol properties

C. H. Song¹, M. E. Park¹, H. J. Ahn¹, K. H. Lee¹,², Y. Lee¹, J. Y. Kim², K. M. Han¹, J. Kim⁴, Y. S. Ghim⁵, and Y. J. Kim¹

¹Dept. of Environmental Science and Engineering, Gwangju Inst. of Science and Technology (GIST), Gwangju, Korea, and also at Advanced Environmental Monitoring Research Center (ADEMRC), Gwangju Inst. of Science and Technology (GIST), Gwangju, Korea
²Hazardous Substance Research Center, Korea Inst. of Science and Technology (KIST), Seoul, Korea
³Earth System Science Interdisciplinary Center (ESSIC), Univ. of Maryland, MD 20742, USA
⁴Dept. of Atmospheric Science, Yonsei Univ., Seoul, Korea
⁵Dept. of Environmental Science, Hankuk Univ. of Foreign Studies, Yongin-si, Gyeonggi-do, Korea

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Correspondence to: C. H. Song (chsong@gist.ac.kr)
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Abstract
In this study, the spatio-temporal and seasonal distributions of EOS/Terra Moderate Resolution Imaging Spectroradiometer (MODIS)-derived aerosol optical depth (AOD) over East Asia were analyzed in conjunction with US EPA Models-3/CMAQ v4.3 modeling. In this study, two MODIS AOD products (τMODIS-M-BAER and τMODIS-NASA) retrieved through a modified Bremen Aerosol Retrieval (M-BAER) algorithm and NASA collection 5 (C005) algorithm were compared with the AOD (τCMAQ) that was calculated from the US EPA Models-3/CMAQ model simulations. In general, the CMAQ-predicted AOD values captured the spatial and temporal variations of the two MODIS AOD products over East Asia reasonable well. Since τMODIS cannot provide information on the aerosol chemical composition in the atmosphere, different aerosol formation characteristics in different regions and different seasons in East Asia cannot be described or identified by τMODIS itself. Therefore, the seasonally and regionally varying aerosol formation and distribution characteristics were investigated by the US EPA Models-3/CMAQ v4.3 model simulations. The contribution of each particulate chemical species to τM-BAER, τNASA, and τCMAQ showed strong spatial, temporal, and seasonal variations. For example, during the summer episode, τM-BAER, τNASA, and τCMAQ were mainly raised due to high concentrations of (NH₄)₂SO₄ over Chinese urban and industrial centers and secondary organic aerosols (SOAs) over the southern parts of China, whereas during the winter episode, τM-BAER, τNASA, and τCMAQ were higher due largely to high levels of NH₃NO₃ formed over the urban and industrial centers, as well as in areas with high NH₃ emissions. In addition, the accuracy of τM-BAER and τNASA was evaluated by a comparison with the AOD (τAERONET) from the AERONET sites in East Asia. Both τM-BAER and τNASA showed a strong correlation with τAERONET around the 1:1 line (R=0.79), indicating promising potential for the application of both the M-BAER and NASA aerosol retrieval algorithms to satellite-based air quality monitoring studies in East Asia.
1 Introduction

Tropospheric aerosols are important components in the atmospheric system. They affect the global radiation budget directly by scattering or absorbing solar radiation and indirectly by changing cloud condensation nuclei (CCN) concentrations and cloud lifetimes in the atmosphere (Twomey et al., 1984; Charlson et al., 1992; Kaufman et al., 2002; Ramanathan et al., 2007). Furthermore, tropospheric aerosols provide important surfaces for heterogeneous reactions between gas-phase air pollutants and atmospheric particles. Through the heterogeneous reactions, acidic and alkaline substances such as sulfate, nitrate, ammonium, and secondary organic aerosols (SOAs) are formed in the tropospheric aerosols, causing urban-, regional-, and global-scale air pollution (Jacobson, 1999; Ramanathan and Crutzen, 2003; Seinfeld et al., 2004).

The generation, transport, and formation of tropospheric aerosols and aerosol climatology (radiative effects) have often been studied by several collaborative research campaigns in East Asia, which integrate ground-based monitoring, aircraft and ship measurements, and 3-dimensional (3-D) chemistry-transport modeling (CTM) efforts (e.g., PEM-West A and B, ACE-Asia, TRACE-P). However, such efforts have been limited by the inability of the point (ground-based monitoring) and line (aircraft- and shipborne measurements) measurements in the campaign activities to completely capture the spatially and temporally varying, regional-scale aerosol generation, formation, and transport characteristics.

For the last couple of decades, satellite-based, remote-sensing techniques have been greatly developed. Particularly, environment observing satellites have provided 2-D, column-integrated aerosol “optical” concentrations as important complementary and/or alternative data to the current point- and line-based “chemical” measurements (King et al., 1999; Singh and Jacob, 2000). The satellite-derived, aerosol optical properties have been applied to various research areas in air pollution studies such as: i) identifying the sources of air pollution (e.g., Herman et al., 1997), ii) estimating ground-level particulate concentrations (e.g., Wang and Christopher, 2003; van Donkelaar et al., 2006), iii) understanding the long-range transport of air pollutants across oceans and continents (e.g., Husar et al., 2001; Colarco et al., 2002; Takemura et al., 2002; Lee et al., 2005b), iv) developing an advanced air quality forecast system (e.g., Al-Saadi et al., 2005), and v) aerosol climatology studies (e.g., Holben et al., 2001; Weaver et al., 2002; Kaufman et al., 2002; Hsu et al., 2003). Currently, many satellite platforms, such as Total Ozone Mapping Spectrometer (TOMS), Sea-Viewing Wide Field-of-View Sensor (SeaWIFS), Moderate Resolution Imaging Spectroradiometer (MODIS), Medium Resolution Imaging Spectroradiometer (MERIS), Ozone Monitoring Instrument (OMI), Multi-angle Imaging Spectroradiometer (MISR), and Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO), have provided various qualities and levels of aerosol optical properties (King et al., 1999; Singh and Jacob, 2000). Meanwhile, remotely sensed, aerosol optical column properties are also provided by ground-based, sunphotometer network and data archive, such as the Aerosol Robotics Network (AERONET) managed by the NASA Goddard Space Center (Holben et al., 1998). Therefore, it would be desirable and promising to use these abundant, satellite- and sunphotometer-derived aerosol optical properties in urban- and regional-scale air pollution studies, in conjunction with the point- and line-based chemical measurements and 3-D photochemical modeling. In this study, we therefore apply the satellite- and sunphotometer-derived aerosol optical depth (hereafter, denoted as AOD or \(\tau\)) to particulate pollution studies over East Asia in the link with 3-D Eulerian CTM simulations.

The main aims of the study were three-fold. The first was to examine the seasonally and regionally varying aerosol characteristics in East Asia using i) the particulate chemical composition from 3-D Eulerian CTM simulations, ii) two MODIS AOD products (\(\tau_{\text{MODIS}}\) and \(\tau_{\text{NASA}}\)) retrieved through a modified Bremen Aerosol Retrieval (M-BAER) algorithm and NASA collection 5 (C005) algorithm, and iii) the AOD product \(\tau_{\text{AERONET}}\) from the NASA AERONET sites in East Asia. The second aim was to compare the \(\tau_{\text{BAER}}\) and \(\tau_{\text{NASA}}\) with \(\tau_{\text{AERONET}}\), and validate the 3-D CTM simulations with 2-D domain-wide satellite-derived \(\tau_{\text{BAER}}\) and \(\tau_{\text{NASA}}\) and point measurement-based \(\tau_{\text{AERONET}}\). The third aim was to test the potential for the application of both the M-
BAER and NASA aerosol retrieval algorithms to satellite-based air quality monitoring and aerosol climate forcing studies in East Asia, by comparing the four different AOD products ($\tau_{\text{BAER}}$, $\tau_{\text{NASA}}$, $\tau_{\text{CMAQ}}$, $\tau_{\text{AERONET}}$) over East Asia.

In particular, the use of the 2-D domain-wide, satellite-derived, aerosol column data for the validation of the 3-D CTM simulations overcomes the typically faced "point-versus-grid errors" and "line-versus-grid errors" (also, often called "pencil-thin effect"), when the 3-D CTM simulation results are compared with data from the ground station, ship, and aircraft measurements. As the 3-D CTM, we used the US EPA Models-3/CMAQ v4.3 (Community Multi-scale Air Quality) model. The AOD products retrieved from the MODIS satellite sensor on board the NASA EOS/Terra satellite were used for the satellite-derived aerosol optical properties.

The integrated analysis of satellite-, AERONET-, and Models-3 CMAQ-derived $\tau$ has many merits. For example, while $\tau_{\text{MODIS}}$ can provide spatial and temporal, 2-D domain-wide, aerosol distributions, it cannot provide information on the chemical composition of the atmospheric aerosols. This limitation can be overcome by using the 3-D CTM simulations. By conducting US EPA Models-3/CMAQ modeling in the link with the satellite-derived aerosol data, different aerosol formation characteristics in different regions and different seasons in East Asia can be more clearly described and identified.

This study carried out a comprehensive investigation by closely integrating the remote sensing data with the 3-D CTM products considering the detailed atmospheric gas/aerosol chemico-physical processes in the 3-D CMAQ modeling. This is a fundamental study to investigate the seasonally and regionally varying, aerosol formation and distribution characteristics in East Asia. Based on the results, a further aim is to establish a state-of-the-science chemical weather forecast system in East Asia in the near future, as in the USA (e.g., Al-Saadi et al., 2005), and to more accurately estimate the seasonally varying direct climate forcing by primary and secondary anthropogenic aerosols and mineral dust in East Asia.

2 Model descriptions

In order to conduct CTM over East Asia, we used a one-way coupling of PSU/NCAR MM5 (Pennsylvania State University/National Center for Atmospheric Research Mesoscale Model 5) to the US EPA Models-3/CMAQ model along with the East Asian emission inventory. The procedures are explained in detail below.

2.1 US EPA Models-3/CMAQ modeling

We performed 3-D Eulerian modeling over East Asia, using the US EPA Models-3/CMAQ v4.3 model (Byun and Ching, 1999; Byun and Schere, 2006). In this modeling study, both Carbon Bond Mechanism 4 (CBM4) and Carnegie-Mellon University (CMU) aqueous-phase chemistry were selected for the full consideration of gas-phase and aqueous-phase chemistries, respectively. For the advection and dry deposition of the gas and particulate species, the Piece-wise Parabolic Method (PPM) and latest version of Wesley scheme were chosen.

With regard to the aerosol chemical, dynamic, and thermodynamic processes, one of the most salient features of the US EPA Models-3/CMAQ v4.3 model is that the aerosol module includes the ISORROPIA aerosol equilibrium model and the mathematically-efficient, modal approach. The performance of the ISORROPIA model in predicting gas-particle distributions of ambient acidic (e.g., H$_2$SO$_4$, HNO$_3$, HCl) and alkaline (e.g., NH$_3$) species is presented in Nenes et al. (1998). Capaldo et al. (2000) also reported that interactions between gas-phase species and fine-mode particles are sufficiently fast for this heterogeneous process to be dealt with in a thermodynamic manner. In contrast, interactions between gas-phase species and coarse-mode particles are so slow that this process must be treated in a kinetic or dynamic manner, in a method called the "CMU hybrid approach". The aerosol modal approach has been applied to 3-D photochemical models because it can provide mathematically-convenient forms of formulas describing various aerosol processes such as coagulation, condensation, and particle growth (Binkowski, 1999; Binkowski and Roselle, 2003).
The study domain covered approximately the region from 100° to 150° E and from 20° to 50° N (shown in Fig. 1), which includes all of Korea, Japan, most of eastern China, and parts of Mongolia and Russia. The region has dramatic variations in topography and land type, and features mixtures of industrial/commercial/urban centers and agricultural/rural regions. Figure 1 illustrates the Chinese urban-industrial centers, such as Bohai Bay, Sichuan Basin, and Yangtze Delta areas, denoted as regions A, B and C, respectively, and the region with high NH₃ emission due to strong agricultural and livestock farming activities (Hebei, Shandong, Henan, Jiangsu, and Anhui provinces). In addition, dust storms frequently erupt from the Gobi desert, loess plateau, and Inner Mongolia, typically in spring.

In the 3-D CMAQ modeling, the horizontal grid spacing was 108 km × 108 km with 46 and 33 grids in the x and y directions, respectively, and vertical domain ranged from 1000 hPa to 180 hPa, with 24 terrains following \( \sigma \)-coordinates, giving a total of 36,432 grid points.

The 3-D photochemical modeling was conducted for four episodes, representing the four seasons, in East Asia. The periods of the four episodes, each approximately 3 weeks long, were: i) 9–27 November 2001 (Late fall); ii) 25 March–13 April 2002 (Spring); iii) 24 August–13 September 2002 (Late Summer); and iv) 11–28 February 2003 (Winter). US EPA Models-3/CMAQ CTM was conducted for the four episodes using the MET fields generated from the PSU/NCAR MM5 modeling.

2.2 Meteorological modeling

As a meteorological preprocessor for the US EPA Models-3/CMAQ CTM, PSU/NCAR MM5 was used in this study. In the meteorological modeling (MET modeling), the 3-D non-hydrostatic primitive equation was selected as the governing equation. The horizontal spacing was 108 km × 108 km, which follows the horizontal spacing of the US EPA Models-3/CMAQ model. However, the size of the domain for MM5 modeling was larger by two grids than that for US EPA Models-3/CMAQ modeling in the 4 compass directions in order to minimize the uncertainties that could occur when meteorological boundary conditions were set for US EPA Models-3/CMAQ modeling. For the same reason, the vertical domain for MET modeling was also higher, from 1000 to 70 hPa, than that for US EPA Models-3/CMAQ modeling, with 30 terrain following \( \sigma \)-coordinates. For the MET modeling, 2.5° × 2.5° re-analyzed, National Centers for Environmental Prediction (re-analyzed NCEP) data were used to drive MM5. The re-analyzed NCEP data used in this study also contained the sea surface temperature (SST) data. In addition, the data included in the PSU/NCAR storage were used to give consideration for the terrain heights within the domain. In order to improve the accuracy of the MET fields, NCEP Automated Data Processing (ADP) global surface and upper air observation data were utilized by employing Four-D Data Assimilation (FDDA) techniques with nudging coefficients of \( 2.5 \times 10^{-4} \) for temperature and wind and of \( 1 \times 10^{-5} \) for mixing ratios. In the MM5 modeling, Blackadar scheme and five-layer land surface models (LSMs) were used for the planetary boundary layer (PBL) parameterization and associated ground temperature scheme, respectively. Rapid Radiative Transfer Model (RRTM), Grell scheme, and Reisner-2 scheme were chosen to give consideration for atmospheric radiation, cumulus parameterization, and cloud microphysics, respectively (Grell et al., 1994; Reisner et al., 1998). Using these schemes, grided MET fields were generated with a high temporal resolution of 1-h interval. After the generation of the grided MET fields from the MM5 modeling, the data were then processed and converted by the meteorological-chemistry interface processor (MCIP) to produce MET inputs for the US EPA Models-3/CMAQ model.

2.3 Emissions

Resolved emission data at 1° × 1° were obtained from the ACE-ASIA (Asian Pacific Regional Aerosol Characterization) emission estimation web site (http://www.cgrer.uiowa.edu/people/carmichael/ACESS/Emissions-data/main.html). The emission data for NOₓ and SO₂ were of primary importance, as they are the precursors of particulate nitrate and sulfate, respectively. The emission inventory for the two primary pollutant species in East Asia has been continuously improved and evaluated in the
literature (van Aardenne et al., 1999; Streets et al., 2003; Woo et al., 2003). Although some studies indicated that the NO\textsubscript{x} emissions would be \( \sim 30\% \) underestimated in East Asia through comparisons between model-predicted and satellite-derived NO\textsubscript{2} column concentrations (Ma et al., 2006), it is generally believed that the uncertainties in their emissions are relatively small. Uncertainty levels of \( \pm 37\% \) and \( \pm 16\% \) were reported by Streets et al. (2003) for NO\textsubscript{x} and SO\textsubscript{2} emissions, respectively, in East Asia. NH\textsubscript{3} emission in East Asia is more uncertain and poorly-understood, despite being an important gas-phase precursor of particulate ammonium. The NH\textsubscript{3} emission used in this study was estimated by taking four major sources into account: i) fertilizer applications, ii) livestock-farming (animal excreta and manures), iii) biomass/bio-fuel burnings, and iv) industrial combustion (Woo et al., 2003). Non-Methane Volatile Organic Compound (NMVOC) emissions are also important, since they are SOA precursors. In this study, we conducted chemical speciation (chemical species splitting) of the total NMVOC emissions in East Asia, using the SPECIATE database built up by US EPA. For the consideration of the biogenic isoprene and terpene emissions, we used the monthly emission data from the Global Emission Inventory Activity (http://weather.engin.umich.edu/geia). Emissions of primary carbonaceous particles (black carbon and organic carbon: BC and OC) were also included in the emission inventory. Some of the aforementioned emissions are shown in Fig. 2.

Spring is a typical dust season in East Asia. During the spring episode studied, high wind speeds over the Gobi desert resulted in significant dust emissions. In order to estimate the mass flux of dust emissions \( (E_{\text{Dust}}) \), a method that employs both friction velocity \( (u_{*}) \) and threshold friction velocity \( (u_{*,t}) \) was utilized (Gillette and Passi, 1988; Gillette et al., 1992):

\[
E_{\text{Dust}} = C A \Delta t \left[ u_{*} \left( 1 - \frac{u_{*,t}}{u_{*}} \right) \right]
\]

where \( C \) is a combined constant, and \( A \) and \( \Delta t \) represent the surface area and time period considered for dust emissions, respectively. This parameterization was successfully tested by several previous studies over the same East Asian domain (Phadnis and Carmichael, 2000; Song and Carmichael, 2001a, b). Unfortunately, however, the sea-salt particle (another natural aerosol) generation over the ocean areas of the domain was not considered in this study, partly because the primary study focus was on the modeling and monitoring investigations for anthropogenic-influenced pollution areas. This consideration should, however, be included in future study.

3 Remote sensing of aerosol optical properties

In this study, we used two different types of satellite-derived, aerosol optical properties: Aerosol Index (AI) and AOD. AI was retrieved from the Earth Probe TOMS platform and AOD from the NASA/Terra MODIS satellite sensor. The AOD values at several ground locations inside the domain were also obtained through a sunphotometer network of AERONET managed by the NASA Goddard Space Flight Center (Holben et al., 1998). These aerosol optical properties were then applied to further analysis in conjunction with 3-D Eulerian MM5-CMAQ modeling.

3.1 TOMS Aerosol Index

TOMS instruments (McPeters et al., 1996) have been providing useful global UV radiation, ozone, and UV absorbing aerosol data for more than two decades. The TOMS AI was initially computed for the correction of aerosol-induced errors in the retrieval of total ozone (Herman et al., 1997; Torres and Bhartia, 1999). The TOMS AI is calculated on the basis of the difference between the measured spectral contrast of the 360 nm and 331 nm wavelength radiances and the contrast calculated from the radiative transfer theory for a pure molecular atmosphere, which allows for the detection of UV absorbing aerosols over both land and ocean. In the current version 8 Nimbus7 TOMS (1979–1993), Earth Probe TOMS (1996–2007) and version 2 Aura OMI
(2004-present) algorithms, it is mathematically defined as:

\[
AI = 100 \left\{ \log_{10} \left( \frac{\rho_{360}}{\rho_{331}} \right)_{\text{meas}} - \log_{10} \left( \frac{\rho_{360}}{\rho_{331}} \right)_{\text{calc}} \right\}
\]

(2)

where the indices meas and calc indicate the radiance measured by TOMS and the radiance calculated for pure Rayleigh scattering, respectively. AI is positive for absorbing aerosols (e.g., dust and BC particles) and negative for non-absorbing aerosols (e.g., sulfates). In this study, the TOMS AI data were retrieved from the Earth Probe TOMS platform.

3.2 MODIS AOD data

The MODIS instrument is mounted on two NASA Earth Observation System (EOS) platforms: the Terra and Aqua satellites. The former is on a descending orbit flying southward across the equator around 10:30 local sun time, while the latter Aqua satellite is on an ascending orbit flying northward around 13:30 local sun time. MODIS has 36 channels ranging from 0.41 \( \mu m \) to 14 \( \mu m \) at three different spatial resolutions (250 m, 500 m and 1 km). In this study, we used the AOD retrieved from the EOS/Terra MODIS platform.

For spatially-resolved aerosol retrieval from MODIS data, a modified version of the BAER algorithm (von Hoyningen et al., 2003; Lee et al., 2005, 2007) was applied. To avoid inhibition from cloud and sun-glint effects, the MODIS L1B data were first processed to filter out cloud and sun-glint pixels (Ackerman et al., 1998; Martins et al., 2002), after which the following radiative transfer equation was used to calculate the aerosol reflectance (\( \rho_{AERO} \)):

\[
\rho_{AERO}(\lambda) = \rho_{TOA}(\lambda) - \rho_{Ray}(\lambda) - \frac{T_{0} \cdot T_{s} \cdot \rho_{Surf}}{1 - s \cdot \rho_{Surf}}
\]

(3)

where \( \rho_{TOA} \), \( \rho_{Ray} \), and \( \rho_{Surf} \) represent the Top-of-the-Atmosphere (TOA) reflectance, Rayleigh path radiance, and surface reflectance, respectively, \( T_{0} \) and \( T_{s} \) the total and surface transmittance, respectively, and \( s \) the hemispheric reflectance. As shown in Fig. 3 and Eq. (3), the MODIS-BAER algorithm is composed of three parts. First, MODIS TOA reflectance in a cloud-free pixel is separated from Rayleigh path radiance. The Rayleigh path radiance can be determined from the altitude-dependent surface pressure \( p(z) \) in each pixel (Bucholtz, 1995). Second, the surface reflectance obtained from the “linear mixing model” is separated (shown in Eq. 4 below). Finally, AOD can be determined by the pre-calculated Look up tables (LUT) approach.

M-BAER was differentiated from the NASA operational, Level-2 Collection 5 (C005) algorithm in the determination method of the surface reflectance. The NASA Level-2 C005 uses “spectral conversion method”, based on the VIS/SWIR ratios \( \left( \rho_{0.47,0.66,\mu m} / \rho_{2.13,\mu m} \right) \). Although this spectral conversion factor is quite dependent on land surface cover type, their assessments of accuracy are reported to be reasonable for dark pixel (Kaufman et al., 1998; Levy et al., 2007). In the M-BAER, the surface reflectance is determined by a linear mixing model of the spectral reflection of “green vegetation” and “BAER soil” to determine the land surface reflectance in BAER (von Hoyningen et al., 2003; Lee et al., 2006).

\[
\rho_{Surf}(\lambda) = C_{veg} \rho_{Veg}(\lambda) + (1 - C_{veg}) \rho_{Soil}(\lambda)
\]

(4)

where \( \rho_{Veg}(\lambda) \) and \( \rho_{Soil}(\lambda) \) are the spectral reflectance of “green vegetation” and “BAER soil”, respectively, and \( C_{veg} \) is the vegetation fraction in each pixel. In Eq. (4), a modified aerosol free vegetation index (AFRI) (Karnieli et al., 2001) was used instead of normalized difference vegetation index (NDVI) as \( C_{veg} \), since the NDVI derived from red and near-infrared (NIR) channels tends to be strongly affected by the presence of aerosols, which could therefore lead to underestimation of vegetation (von Hoyningen-Huene et al., 2003).

Lee et al. (2006, 2007) have applied the M-BAER algorithm to the retrieval of \( \tau \) from the NASA EOS/Terra MODIS platform. They then evaluated the performance of the M-BAER algorithm over “South Korea” by comparing \( \tau_{M-BAER} \) with \( \tau_{AERONET} \). They found that the M-BAER algorithm worked very well, showing good agreement between
This study is an extension of their works into “East Asia” in regard to satellite-derived aerosol optical property monitoring. In addition, this study compared two aerosol products (τ_{M-BAER} and τ_{NASA}) and evaluate the accuracy of both the satellite-derived AOD with τ_{CMAQ} and τ_{AERONET} in East Asia.

3.3 AERONET data

Two sunphotometer/skyradiometer networks have been operated in East Asia: AERONET and SKYNET (Skyradiometer Network). The former is a federated sunphotometer network and data archive for aerosol characterization that has been managed by the NASA Goddard Space Flight Center in USA (Holben et al., 1998). The latter is a skyradiometer (an instrument similar to sunphotometer) network established in late 1997 for studying the role of aerosols in climate change as a part of GEWEX Asian Monsoon Experiment/Asian Automatic Weather Station Network (GAME/AAN) (Kim et al., 2004). In this study, we obtained the sunphotometer-derived AOD (τ_{AERONET}) from AERONET.

AERONET has 21 ground-based monitoring stations within our East Asian modeling domain (this will be shown in Fig. 10), and provides total column spectral τ, single scattering albedo (ω_o), and reflective index at visible and NIR wavelengths. Among the aerosol properties, the spectral τ provided by AERONET are regarded as “ground true values”, since they are unaffected by surface reflectance, unlike the satellite-derived τ.

In this study, we selected τ at 550 nm, by using a power law interpolation. For example, at the Gosan AERONET site the AOD values at 550 nm were interpolated between τ at 440 nm and 675 nm. In addition, the sunphotometer data closest to the satellite scanning time were taken from the AERONET sites and τ_{MODIS} were selected at the nearest locations to the AERONET sites for further comparison studies.

4 Results and discussions

Spatial, temporal, and seasonal distributions of aerosol mass loading and τ generated by US EPA Models-3/CMAQ modeling were compared with MODIS-derived AOD and AOD from AERONET over East Asia. Following these comparisons, we discuss the species-wise formation characteristics of particulate pollutants in the different regions of East Asia for the four seasons.

4.1 CMAQ- and MODIS-derived AOD in East Asia

4.1.1 Estimation of CMAQ-derived AOD

Section 3.2 presented the method to retrieve τ_{MODIS} from the NASA EOS/Terra MODIS platform via the M-BAER and NASA C005 algorithms. This section introduces methods to estimate τ_{CMAQ} from the particulate concentrations generated from the US EPA Models-3/CMAQ model. AOD (τ) is theoretically calculated by integrating the aerosol extinction coefficient (σ_{ext}(z)) with respect to altitudes (z), i.e.:

\[ \tau = \int \sigma_{ext}(z) \cdot dz \]  

The method by which σ_{ext}(z) is calculated from the particulate composition is therefore important. Actually, there are two such methods. The first is based on the Mie theory, in which the extinction coefficient of a certain species \( i \) (\( \sigma_{ext,i} \)) is calculated as a function of density of particulate species \( i \) (\( \rho_i \)), mass of particulate species \( i \) at an altitude \( z \) (\( M_i(z) \)), extinction efficiency of particulate species \( i(Q_{ext,i}) \), and effective radius of particulate species \( i(r_{eff,i}) \) (d’Almeida et al., 1991; Chin et al., 2002):

\[ \sigma_{ext,i}(z) = \frac{3Q_{ext,i}M_i(z)}{4\rho_i r_{eff,i}(z)} \]
The total aerosol extinction coefficient \( \sigma_{\text{ext}} \) is then calculated by \( \sigma_{\text{ext}} = \sum \sigma_{\text{ext}, i} \), based on the assumption that all the particulate species, except for particulate water, are externally mixed. Of the aerosol properties in Eq. (6), \( r_{\text{eff}, i}(z) \) is a function of the size-distribution of particulate species \( i \) and altitude \( z \), and \( Q_{\text{ext}, i} \) is a function of \( r_{\text{eff}, i} \), refractive index \( (m_i) \) of particulate species \( i \), and wavelength \( (\lambda) \), i.e., \( Q_{\text{ext}, i}(r_{\text{eff}, i}, m_i, \lambda) \). Also, the following properties are strongly variable with ambient relative humidity (RH), since each particulate species has different hygroscopic capability. Thus, all the aerosol properties in Eq. (6) are dependent on RH (d’Almeida et al., 1991; Chin et al., 2002; and more references therein). The Mie-theory-based extinction coefficient calculations are, therefore, based on several assumptions, as presented by Chin et al. (2002; 2004). For example, in Chin et al. (2002, 2004)’s global 3-D modeling studies, the size-distribution of each particulate species was assumed (or determined), and then fixed during the 3-D model simulations. However, for example, the size-distributions of dust particles have high spatial and temporal variations due to their different emission strength and deposition characteristics, and sulfate and SOA distributions are also spatially and temporally variable due to different particle growth rates. In addition, the changes of the aerosol properties due to hygroscopicity are highly uncertain. In East Asia, the consideration of \( \text{NH}_4\text{NO}_3 \) is of primary importance, but its size-distribution has not been sufficiently characterized to enable the use of the Mie-theory-based extinction coefficient calculations. Therefore, following consideration of these uncertainties we estimated \( \sigma_{\text{ext}} \) with an empirical correlation, often called the “reconstructed extinction coefficient method”. The method was first proposed by Malm et al. (1994) to investigate the spatial and temporal variability of haze and visibility impairment in USA, and has been continuously improved by several researchers (e.g., Malm and Kreidenweis, 1997; Malm, 2000). Among the correlations available, we adopt the latest version of the formula proposed by Malm (2000) as a part of the Integracy Monitoring of Projected Visual environment (IMPROVE) program:

\[
\sigma_{\text{ext}}(\text{Mm}^{-1}) = 3.0 \times f(RH) \times \left\{ \left[ (\text{NH}_4\text{SO}_4) + [\text{NH}_4\text{NO}_3] \right] + 4.0 \times [\text{SOAs}] + 10.0 \times [\text{BC}] + 1.0 \times [\text{fine - dust}] + 0.6 \times [\text{coarse - dust}] \right\}
\]

(7)

This method is also based on the same assumption as the Mie-theory-based method, i.e., all the particulate species, except for particulate water, are externally mixed. As shown in Eq. (7), \( \sigma_{\text{ext}} \) is conveniently calculated from the particulate concentrations (in \( \mu g \text{ m}^{-3} \)) of \( (\text{NH}_4\text{SO}_4), \text{NH}_4\text{NO}_3, \text{BC}, \text{SOAs}, \) and fine-mode and coarse-mode dust, which are generated from the US EPA Models-3/CMAQ model. In Eq. (7), the variability in \( \sigma_{\text{ext}} \) caused by variable RH was considered by hygroscopic growth factor or extinction enhancement factor, \( f(RH) \). In the calculations, only \( (\text{NH}_4\text{SO}_4) \) and \( \text{NH}_4\text{NO}_3 \) were considered hygroscopic. The consideration of \( f(RH) \) is of primary importance, since \( \sigma_{\text{ext}} \) varies greatly with RH, particularly when RH is larger than 80%. In addition, \( f(RH) \) exhibits seasonal variation which is accounted for by using the following equation:

\[
f(RH) = b_0 + b_1 \left( \frac{1}{1 - RH} \right) + b_2 \left( \frac{1}{1 - RH} \right)^2
\]

(8)

The correlation parameters \( (b_0, b_1, \) and \( b_2) \) in Eq. (8) for the different seasons are presented in Table 1. Here, a further assumption made in using Eq. (6) is that the aerosol hygroscopic characteristics in East Asia are the same as those in USA. In addition, for the calculations of RH (= \( \sigma(z)/\sigma_w(z) \)), water vapor pressure \( \sigma(z) \) is calculated by the equation proposed by Mattis et al. (2002) and saturation vapor pressure is estimated by the Bolton equation shown below:

\[
\theta_w(z) = 6.112 \times \exp \left( 17.67 \times \frac{T - 273.15}{T - 29.65} \right)
\]

(9)

where \( T \) is temperature in K and \( \theta_w(z) \) indicates the saturation vapor pressure at an altitude \( z \).
4.1.2 CMAQ-derived AOD vs. MODIS-derived AOD

The two MODIS-derived AOD ($\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$) were compared with both the aerosol column mass loading (mg m$^{-2}$) and the $\tau_{\text{CMAQ}}$ values calculated from the aerosol composition produced by the Models-3/CMAQ model. $\tau_{\text{CMAQ}}$ was calculated based on the reconstructed extinction coefficient method (Sect. 4.1.1). In Fig. 4, the spatial distributions of episode-average $\tau_{\text{M-BAER}}$, $\tau_{\text{NASA}}$, $\tau_{\text{CMAQ}}$, and aerosol column mass loading are shown for the four seasonal episodes. The high $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ values appeared over four highly polluted areas: (i) Bohai Bay, (ii) Sichuan Basin, (iii) lower course of the Yangtze river (also called “Yangtze River Delta”), and (iv) high NH$_3$ emission area (refer to Fig. 1). During the spring episode, a large dust storm broke out, and was transported over long distances through Manchuria and the Korean peninsula. Therefore, Fig. 4 shows a high aerosol mass loading over these areas during this episode (note that the scale of the column aerosol mass loading in spring was changed to better present the dust plume in Fig. 4). $\tau_{\text{CMAQ}}$ along the transport track of the dust storm was not noticeably high, although it ranged between 0.5 and 1.0 over Manchuria and the northern parts of the Korean peninsula, possibly due to the lower sensitivity of the reconstructed extinction coefficient method to the coarse-mode dust concentrations, as shown in Eq. (7). Figure 4 shows high $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ values over the ocean but not over the land. In addition, there was a clear discontinuity in $\tau_{\text{MODIS}}$ over land and ocean in the dust plume, i.e., high $\tau_{\text{MODIS}}$ over the Bohai Sea and the Sea of Japan (also known as the East Sea), but low $\tau_{\text{MODIS}}$ over Manchuria and the Korean peninsula. This may have two causes: i) the different aerosol retrieval methods in the M-BAER and NASA algorithms over the land and ocean (refer to Sect. 3.2) and ii) the cloud pixel screening procedure in the AOD retrieval algorithms. Regarding the latter, dust plumes are typically transported behind or below the cold frontal clouds in East Asia. During the spring episode, a dust storm erupted over the eastern parts of Inner Mongolia on 7 April 2002, and was transported out over Sakhalin and Hokkai Islands, passing through Manchuria and the Korean peninsula on 8 and 9 April 2002.

Figure 5 shows snapshots of the infrared difference dust index (IDDI) with clouds from the NOAA satellite on 7 April to 9 April 2002. As shown in Fig. 5, the dust plumes were accompanied by clouds. Therefore, the many pixel values reflecting high dust concentrations were screened out. These procedures resulted in low average $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ over Manchuria, as shown in Fig. 4. Further analysis of the dust plumes with TOMS Al is discussed in Sect. 4.1.3.

In addition, no $\tau_{\text{MODIS}}$ values were reported over the desert and loess areas due to surface glint effects (cf. Fig. 1). The surface glint areas (white-colored areas in Fig. 4) were the largest in winter and many parts of the desert and loess areas were also covered by snow and ice during the winter season. In contrast, the grass-covered areas were the largest in summer, and thus the surface glint areas were shrunk during the summer episode.

Although the integrated satellite-CMAQ model analysis has many merits, further “chemical” analysis is still useful. $\tau_{\text{MODIS}}$ can provide the 2-D domain-wide, spatial distributions of columnar aerosol optical properties, but not the species-wise, spatially-resolved, aerosol chemical composition distributions. Therefore, the different aerosol formation characteristics in different regions and different seasons in East Asia could not be described or identified by $\tau_{\text{MODIS}}$ itself, and the analysis requires comprehensive CTM such as US EPA Models-3/CMAQ modeling in conjunction with the satellite-derived, aerosol optical properties. This is discussed in Sect. 4.1.4.

4.1.3 Dust event and TOMS Al

As discussed in Sect. 3.1, Al is a good optical indicator for absorbing aerosols (such as dust and BC), especially when they are transported through the “free” troposphere. Based on this, many scientists have investigated the ability of remotely sensed, high Al plumes concurrently occurring over the same regions to represent dust plumes transported over long-distances across the oceans and continents (Phadnis and Carmichael, 2000; Israelevich et al., 2002; Ginoux et al., 2001; Ginoux and Torres, 2003; Ginoux et al., 2004).
Snapshots for the spatial and temporal Al distributions during the dust period (7 to 9 April 2002) over East Asia are shown in the top three panels in Fig. 6. The transport of model-predicted, high column mass loading (mostly composed of dust particles) through Manchuria and the Korean peninsula are also presented in the bottom three panels in Fig. 6. Figure 6 illustrates the good agreements in the spatial and temporal distributions of both the high Al plumes and the high column mass loading (although both quantities are not directly compared to each other).

In the top panels in Fig. 6, another high Al plume can be found over the Guangxi and Guangdong provinces (near Hong Kong). The high Al plume may have been caused by the BC plume emitted from biomass burning events. Spring is not only the typical dust season, but also the typical biomass burning season in East Asia (Ma et al., 2003; Song et al., 2005b; more references therein). The former usually influences the northeast Asian atmosphere, whereas the latter affects southeast Asian atmosphere, and frequently occurs in the southwestern edges of China, Thailand, Myanmar, and northern Vietnam. Since the biomass burning events took place outside the modeling domain, and as the CMAQ modeling does/can not consider the biomass burning event in the boundary layer conditions, the BC plumes over the Guangxi and Guangdong provinces were not captured in the results of the CMAQ modeling (bottom panels in Fig. 6).

4.1.4 Seasonal and regional formation characteristics of particulate pollutants

Figure 7 presents the spatial distributions of the column concentrations of five major particulate constituents of PM$_2.5$: (i) nitrate (NO$_3^-$), (ii) sulfate (SO$_4^{2-}$), (iii) dust, (iv) BC, and (v) SOAs. Although the spatial distribution of ammonium (NH$_4^+$) is not shown in Fig. 7, it can be estimated from the column concentrations of nitrate and sulfate, because ammonium is present in fine-mode particles as forms of crystallized salt like NH$_4$NO$_3$ and (NH$_4$)$_2$SO$_4$, and/or is neutralized by nitrate and sulfate ions in the particulate water. The NH$_4$NO$_3$ salt formation or NH$_4^+$-NO$_3^-$ ion association takes place by

the following equilibrium reaction (Seinfeld and Pandis, 1998; Jacobson, 1999):

$$\text{NH}_3 (g) + \text{HNO}_3 (g) \rightarrow \text{NH}_4\text{NO}_3 (s) \text{ and/or } \text{NH}_4^+(aq) + \text{NO}_3^-(aq)$$

(R1)

The equilibrium reaction of (R1) is governed by thermodynamic relationships. If the product of partial pressures of gas-phase NH$_3$ and HNO$_3$ is larger than the equilibrium constant ($K_{eq}$), the equilibrium reactions proceed in the forward direction (i.e., the formation or association of NH$_4$NO$_3$ and/or NH$_4^+$-NO$_3^-$ ions). On the contrary, if the product is smaller than $K_{eq}$, the equilibrium is governed by the reverse reaction. The equilibrium is also controlled by temperature, since $K_{eq}$ is a function of temperature. When temperature is low, the equilibrium tends to shift toward the particulate NH$_4$NO$_3$ and/or NH$_4^+$-NO$_3^-$ formation, whereas, if temperature is high, it proceeds in the reverse direction. This heterogeneous process is considered by the ISORROPIA module in the Models-3/CAMQ model (Nenes et al., 1998), and both forward and reverse reactions take place so rapidly that one can treat these processes in a thermodynamic manner, as described in Sect. 2.1 (Capaldo et al., 2000). In contrast, (NH$_4$)$_2$SO$_4$ formation and/or NH$_4^+$-SO$_4^{2-}$ association take place in an "irreversible way":

$$2\text{NH}_3(g) + \text{SO}_4^{2-}(aq) + 2\text{H}^+ \rightarrow (\text{NH}_4)_2\text{SO}_4 (s) \text{ and/or } 2\text{NH}_4^+(aq) + \text{SO}_4^{2-}(aq)$$

(R2)

Once sulfate is formed via SO$_2$+OH reaction and subsequent gas-to-particle conversion, it is then neutralized by gas-phase NH$_3$, following (R2).

Some amounts of nitrate and sulfate can also be formed in fine- and coarse-mode dust particles, in which nitrate and sulfate mainly exist as forms of Ca(NO$_3$)$_2$ and CaSO$_4$. Song et al. (2005a, 2007), however, revealed that major parts (>95% on the mass basis) of NH$_4^+$-NO$_3^-$-SO$_4^{2-}$ are mainly present in anthropogenic pollution particles and that they are almost externally mixed with chemically near-fresh dust particles in the East Asian atmosphere. Only small amounts of nitrate and sulfate can be formed in fine- and coarse-mode dust particles due to the small magnitudes of uptake coefficients of gas-phase sulfate and nitrate precursors (such as SO$_2$, H$_2$SO$_4$, NO$_3$, NO$_2$, N$_2$O$_5$, and HNO$_3$) onto East Asian mineral dust. Based on the conclusions of Song et
al. (2005a, 2007), we neglect the nitrate and sulfate formation in/on dust particles in this study.

The first three columns in Fig. 7 present the spatial distributions of the nitrate, sulfate, and dust concentrations. As discussed previously, dust concentrations are large only in spring due to the dust storm. In Fig. 7, two facts are particularly noteworthy regarding the particulate NH$_4^+$-NO$_3^-$-SO$_4^{2-}$ formation. First, the levels of nitrate (associated with ammonium) were almost comparable to those of sulfate (associated with ammonium). Thus, the NH$_4^+$-NO$_3^-$ formation can/should not be neglected in the CTM studies, particularly over East Asia. Second, the NH$_4^+$-NO$_3^-$ formation characteristics differed greatly from the NH$_4^+$-SO$_4^{2-}$ formation characteristics throughout the four seasons. Regarding the sulfate formation, the SO$_2$ into sulfate conversion rate peaked in summer, because of the high temperature and high levels of hydroxyl radicals (OH) due to the intense solar radiation and high levels of H$_2$O. Therefore, sulfate concentrations were the largest in summer and the smallest in winter. On the contrary, the fine-mode nitrate formation was active at low temperature. Thus, the levels of nitrate were the highest in winter and the lowest in summer. Both fine-mode sulfate and nitrate are the key particulate species that actually affect AOD ($\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$) and therefore climate forcing by aerosols in East Asia. The contributions of both species to $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ increased with increasing RH, because both species are “hygroscopic”, as described in Eq. (7).

In addition, as mentioned above, the contributions of both species to $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ exhibited seasonal variations. In East Asia, (NH$_4$)$_2$SO$_4$ was the main contributor to large $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ in summer, whereas NH$_4$NO$_3$ was in winter. Meanwhile, the extent of the contributions of both species also exhibited regional variations. NH$_4$NO$_3$ was usually formed over high NH$_3$ emission areas (refer to Fig. 1), whereas (NH$_4$)$_2$SO$_4$ was formed mainly over the urban-polluted regions (regions A, B, and C in Fig. 1), since SO$_2$ was mainly emitted from such areas.

BC is a light-absorbing particle. With the same mass concentration, its contribution to $\sigma_{\text{ext}}$ and $\tau$ is large. Generally, the absolute BC levels are usually lower in East Asia than those of nitrate and sulfate. Since BC has similar emission sources to SO$_2$, its spatial distributions are more similar to those of sulfate than those of nitrate.

The contributions of SOAs to $\sigma_{\text{ext}}$ and $\tau$ were also considered to be large, as shown in Eq. (7). The absolute levels of SOAs in East Asia were approximately 0.25–0.33 times lower than those of sulfate and nitrate. The SOA formation is mainly governed by three factors: (i) temperature, (ii) the intensity of solar radiation, and (iii) the levels of precursors (i.e., NMVOCs). The SOA precursors react with OH radicals. The levels of OH radicals are primarily controlled by the intensity of solar radiation, (R3), and are also limited by the levels of H$_2$O. (R4):

\[
\text{O}_3 + h\nu \rightarrow \text{O}^1\text{D} + \text{O}_2 \quad (R3)
\]

\[
\text{O}^1\text{D} + \text{H}_2\text{O} \rightarrow 2\text{OH} \quad (R4)
\]

\[
\text{NMVOCs} + \text{OH(O}_3\text{andNO}_3) \rightarrow \text{SOAs (multi – steps)} \quad (R5)
\]

The subsequent reaction of (R4) proceeds very fast, producing OH radicals in the atmosphere. The reaction rate of (R5) is controlled by the levels of OH radicals and increases with increasing temperature. Therefore, in general SOAs were actively produced in summer. There are other important routes for SOA production. NMVOCs react with other oxidants in the atmosphere, such as O$_3$ and NO$_3$. The concentrations of these two species are highest in summer and lowest in winter. In addition to the high levels of OH, O$_3$ and NO$_3$, biogenic NMVOC emissions were also very active in summer. As mentioned in Sect. 2.3, we used the monthly biogenic isoprene and monoterpane emissions from the GEIA inventory in the US EPA Models-3/CMAQ modeling study. The biogenic emissions were highly active in the southern parts of the Yangtze river. Thus, high levels of SOAs were found in the southern areas of the Yangtze river, and were the key particulate component that actually affected $\sigma_{\text{ext}}$, and thus $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$, over such regions in summer (refer to the last column in Fig. 7).
4.1.5 Statistical analysis

This section presents an investigation into the correlation between \( \tau_{\text{MODIS}} \) and \( \tau_{\text{CMAQ}} \) over a pollution-affected window in the domain (100°E–130°E; 20°N–45°N). Remote ocean and continental areas were excluded from the analysis because we were primarily interested in the pollution-affected areas and did not consider the generation of sea-salt over the ocean. Figure 8 shows reasonably good correlations between \( \tau_{\text{MODIS}} \) and \( \tau_{\text{CMAQ}} \) with correlation coefficients (R) ranging from 0.45 to 0.58. As shown in Fig. 8, the correlations between \( \tau_{\text{M-BAER}} \) and \( \tau_{\text{CMAQ}} \) (blue dots) and between \( \tau_{\text{NASA}} \) and \( \tau_{\text{CMAQ}} \) (red circles) were similar around the 1:1 line. This was further confirmed by statistical analysis.

For further statistical investigations (error and bias analyses), four statistical parameters were introduced: (i) Root Mean Square Error (RMSE), (ii) Mean Normalized Gross Error (MNGE), (iii) Mean Bias (MB), and (iv) Mean Normalized Bias (MNB).

\[
\text{RMSE} = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (\tau_{\text{CMAQ}} - \tau_{\text{MODIS}})^2}
\]

(10)

\[
\text{MNGE} = \frac{1}{N} \sum_{i=1}^{N} \left( \frac{|\tau_{\text{CMAQ}} - \tau_{\text{MODIS}}|}{\tau_{\text{MODIS}}} \right) \times 100
\]

(11)

\[
\text{MB} = \frac{1}{N} \sum_{i=1}^{N} (\tau_{\text{CMAQ}} - \tau_{\text{MODIS}})
\]

(12)

\[
\text{MNB} = \frac{1}{N} \sum_{i=1}^{N} \left( \frac{\tau_{\text{CMAQ}} - \tau_{\text{MODIS}}}{\tau_{\text{MODIS}}} \right) \times 100
\]

(13)

The results from the statistical analyses are presented in Table 2 for the four seasons.

In error analysis, the RMSEs (absolute errors) ranged between 0.23 and 0.30, and the MNGEs (relative errors) between 39.70% (spring) and 70.51% (fall). In the bias analysis, the MBs ranged between \(-0.15 \) (spring) and 0.10 (fall), and the MNBs between \(-19.86\% \) (spring) and 40.34% (summer). Again, as shown in Table 2, the magnitudes of the four statistical values between \( \tau_{\text{M-BAER}} \) and \( \tau_{\text{CMAQ}} \) and between \( \tau_{\text{NASA}} \) and \( \tau_{\text{CMAQ}} \) are similar, indicating that the M-BAER algorithm has equivalent capability to the NASA C005 algorithm in producing \( \tau \).

As discussed previously, \( \tau_{\text{MODIS}} \) tends to be underestimated over the land areas affected by dust plumes due to the cloud pixel screen-out procedure, which causes MB and MNB to have negative values in spring (\(-0.10 \) and \(-11.26\% \) for \( \tau_{\text{M-BAER}} \), \(-0.15 \) and \(-19.86\% \) for \( \tau_{\text{NASA}} \), respectively). However, except for the spring episode, the values of MBs and MNBs were positive, indicating that \( \tau_{\text{CMAQ}} \) was larger than \( \tau_{\text{MODIS}} \). This may have been partly due to the fact that the reconstructed extinction coefficient-based method reacts sensitively to RH, particularly at RH >80%. In addition, there are, of course, uncertainties in the M-BAER and NASA algorithms. Such results are contradictory to those reported by Chin et al. (2004) over East Asia during the ACE-Asia campaign period. Although they used different methods from those in this study, i.e., (i) \( \tau_{\text{NASA}} \) obtained from the NASA Level-2 C004 products and (ii) Mie-theory-based \( \sigma_{\text{ext}} \) calculations, \( \tau_{\text{NASA}} \) tended to be larger than \( \tau_{\text{GOCART}} \) (i.e., MB and MNB would be negative, if Eqs. (12) and (13) were employed in their study; here, GOCART is the name of the 3-D global CTM used by Chin et al., 2004). They interpreted these results by an explanation that \( \tau_{\text{NASA}} \) from the NASA Level-2 C004 products tends to overestimate AOD, mainly due to the possible underestimation of the influences of the surface reflectance (Remer et al., 2005). Based on this, NASA released NASA Level-2 Collection 5 (C005) products processed with an improved surface reflectance consideration (Levy et al., 2007). In addition, Chin et al.’ work (2004) did not consider the particulate \( \text{NH}_4\text{NO}_3 \) formation, and thereby they also omitted the contribution of particulate \( \text{NH}_4\text{NO}_3 \) to \( \sigma_{\text{ext}} \) and \( \tau \). However, as shown in Fig. 7, the contribution of fine-mode \( \text{NH}_4\text{NO}_3 \) to \( \sigma_{\text{ext}} \) and \( \tau \) cannot be neglected in East Asia, and it should be taken into account in the 3-D CTM study over East Asia. Such omission of the \( \text{NH}_4\text{NO}_3 \)
formation would lead to the under-predictions of $\tau_{\text{GOCART}}$. The distribution of $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ can also be compared by probability distribution function (PDF). The PDFs of $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ over the pollution-affected area, presented in Fig. 8, display highly similar shapes for the four seasons.

### 4.2 CMAQ-, MODIS-derived AOD vs. AERONET AOD

This section compares four different kinds of AOD ($\tau_{\text{CMAQ}}$, $\tau_{\text{M-BAER}}$, $\tau_{\text{NASA}}$, and $\tau_{\text{AERONET}}$) at several AERONET sites in East Asia. Particularly, $\tau_{\text{AERONET}}$ has been regarded as the "ground true value", since it is not interfered by surface reflectance (the largest uncertainty source in the retrieval of satellite-derived $\tau$). Therefore, the comparison studies between $\tau_{\text{MODIS}}$ and $\tau_{\text{AERONET}}$ may be able to provide a good opportunity to evaluate the accuracy of the M-BAER and NASA algorithms and US EPA Models-3/CMAQ modeling over East Asia.

AERONET has 21 ground-based monitoring stations in East Asia (Fig. 10) which provide total column spectral $\tau$ visible and NIR wavelengths. As discussed previously, we selected $\tau$ at 550 nm for these comparison studies. The comparisons among the four different types of $\tau$ are shown in Fig. 11 (upper panel of each figure). The selections of the AERONET sites were based on the AERONET data availability in each season (i.e., in each season, the four sites with the largest number of $\tau_{\text{AERONET}}$ were chosen). The particulate "chemical" composition generated from US EPA Models-3/CMAQ modeling is also plotted in Fig. 11 (bottom panel of each figure), which exhibits the good agreement among $\tau_{\text{CMAQ}}$, $\tau_{\text{M-BAER}}$, $\tau_{\text{NASA}}$, and $\tau_{\text{AERONET}}$. This issue will be further analyzed below.

During the spring episode, the peaks at the Gosan and Shirahama sites on 7 to 9 April 2002 were predominantly affected by the dust storm, and the influence lasted till 13 April 2002. During the non-dust period, however, $\tau_{\text{CMAQ}}$, $\tau_{\text{M-BAER}}$, $\tau_{\text{NASA}}$, and $\tau_{\text{AERONET}}$ appeared to be mainly influenced by secondary anthropogenic inorganic compounds such as ammonium, nitrate, and sulfate, as discussed in Sect. 4.1.4, whereas the contribution of SOAs was relatively small. Again, the contribution of sea-salt particles to $\tau$ may have been important, since the two sites were located in coastal areas. However, in US EPA Models-3/CMAQ modeling, the sea-salt emissions were not considered. Unlike the Gosan and Shirahama sites, the influence of dust particles is negligible in Taipei, and $\tau$ is largely impacted by secondary inorganic compounds.

At the Dalanzadgad site, $\tau_{\text{MODIS}}$ was hardly observed, since Dalanzadgad is located inside the Gobi desert (i.e., due to the surface glint effects over the desert areas). $\tau_{\text{AERONET}}$ available at the Dalanzadgad site showed two salient peaks on 3 and 5 April 2002, but they were not predicted by the dust generation model (Eq. 1). In practice, predictions and forecasts of dust storms have been difficult and very challenging (Gillette and Passi, 1988; Gillette et al., 1992; Park and Lee, 2004; Uno et al., 2006). The dust generation model used in this study can predict large- or regional-scale dust storms like the one that erupted on 7 to 9 April 2002 over Manchuria, but not the local-scale dust events that take place inside the Gobi desert and loess plateau areas.

Unlike the spring season, several $\tau_{\text{MODIS}}$ were retrieved at the Dalanzadgad site in summer due to the spreading of grass-covered areas in this area (also refer to Fig. 4). Nevertheless, the magnitude of $\tau$ in Dalanzadgad was small, due to its location in a remote continental background area. In contrast, $\tau$ in Beijing was large and appeared to be primarily affected by ($\text{NH}_4$)$_2$SO$_4$ and secondarily by NH$_4$NO$_3$ in summer (note that in Fig. 11 (summer), the y-axis in Beijing has been re-scaled). Gosan was also greatly affected by ($\text{NH}_4$)$_2$SO$_4$ and NH$_4$NO$_3$ on 9 to 13 September 2002. The magnitude of $\tau$ in winter was also large, but the contributions of particulate chemical species were changed. As shown in Fig. 11 (winter), $\tau$ from Beijing to Shirahama was more affected by the formation of NH$_4$NO$_3$ than of ($\text{NH}_4$)$_2$SO$_4$. In fall, not many $\tau_{\text{AERONET}}$ were available except at the Anmyon site, where both ($\text{NH}_4$)$_2$SO$_4$ and NH$_4$NO$_3$ almost equally contributed to $\tau$.

As mentioned above, we compared $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ with $\tau_{\text{AERONET}}$ at several AERONET sites in East Asia to evaluate the performance of the M-BAER and NASA algorithms in East Asia. First, the correlations between $\tau_{\text{MODIS}}$ and $\tau_{\text{AERONET}}$ were analyzed in Fig. 12 for the four seasons. The two quantities were highly correlated to...
each other around the 1:1 lines with correlation coefficients (R) ranging from 0.72 to 0.96 (except for R between $\tau_{\text{M-BAER}}$ and $\tau_{\text{AERONET}}$ in the fall). The low R value in the fall season was attributed to the availability of $\tau_{\text{AERONET}}$. As discussed previously, not many $\tau_{\text{AERONET}}$ were available in the fall episode. As previously discussed, since 2006 NASA has released new satellite aerosol products, denoted as NASA Level-2 Collection 5 (C005), which are reported to have reduced the levels of $\tau_{\text{NASA}}$ compared with $\tau_{\text{AERONET}}$ (Dobobik et al., 2002; Remer et al., 2005; Levy et al., 2007). As shown in Fig. 12, with reference to $\tau_{\text{AERONET}}$ (ground true value), both $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ (NASA C005 product) were highly correlated around the 1:1 lines during the four seasons. This suggests that: i) both M-BAER and NASA algorithms can produce a high quality $\tau_{\text{MODIS}}$ and ii) the accuracy of $\tau_{\text{M-BAER}}$ is equivalent to that of $\tau_{\text{NASA}}$. This was further confirmed by statistical analysis.

RMSE, MNGE, MB, and MNB among $\tau_{\text{M-BAER}}$ and $\tau_{\text{MODIS}}$, $\tau_{\text{CMAQ}}$, and $\tau_{\text{AERONET}}$ were analyzed in Table 3. Although the largest differences appeared in the fall episode, $\tau_{\text{CMAQ}}$, $\tau_{\text{M-BAER}}$, and $\tau_{\text{NASA}}$ and $\tau_{\text{AERONET}}$ exhibited good agreements in the other three seasons. Again, the large differences in fall were partly caused by the scarcity of $\tau_{\text{AERONET}}$ during the fall episode. Importantly, no significant positive biases ($\tau_{\text{MODIS}} > \tau_{\text{AERONET}}$) were reported, which have been typical in the analyses of $\tau_{\text{MODIS}}$ vs. $\tau_{\text{AERONET}}$ (Chin et al., 2004; Remer et al., 2005; Levy et al., 2007). This may indicate that both $\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$ can capture the spatial and seasonal aerosol characteristics in East Asia better than $\tau_{\text{NASA}}$ (C004).

5 Summary and conclusions

The spatio-temporal and seasonal distributions of $\tau_{\text{MODIS}}$ ($\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$) were compared with those of aerosol column mass loading ($M_{\text{column}}$) and $\tau_{\text{CMAQ}}$ for the four seasonal episodes in East Asia. Although $\tau_{\text{MODIS}}$ can provide information on spatial aerosol mass distributions, they can not provide aerosol chemical composition. Therefore, in order to estimate the aerosol chemical composition over East Asia, the US EPA Models-3/CMAQ v4.3 model was utilized, together with the PSU/NCAR MM5 meteorological model and the ACE-Asia/TRACE-P official emission inventory for East Asia. In this study, $\tau_{\text{CMAQ}}$ was calculated using a reconstructed extinction coefficient-based method, whereas $\tau_{\text{MODIS}}$ was retrieved from MODIS Level-1B (L1B) radiance data using a modified Bremen Aerosol Retrieval Algorithm (M-BAER algorithm) and NASA Collection 5 (C005) algorithm. In addition, $\tau_{\text{AERONET}}$ was retrieved from several AERONET sites in East Asia obtained and used as the "ground true value" in this study.

Both $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ showed high values around Chinese urban and industrial centers, such as Sichuan Basin, Bohai Bay and Yangtze Delta areas, as well as over active agriculture and livestock farming areas due to their high NH$_3$ emissions. For the four season episodes selected in this study, the CMAQ model generated similar levels of $\tau_{\text{CMAQ}}$ to those of $\tau_{\text{MODIS}}$ throughout the domain, with R ranging between 0.45 and 0.68, RMSE between 0.23 and 0.30, MNGE between 39.7% and 70.5%, MB between −0.15 and 0.10, and MNB between −19.9% and 40.3%. In particular, $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ exhibited good agreement in the probability distributions over the domain. However, during the spring episode, $\tau_{\text{CMAQ}}$ was significantly under-predicted compared to $\tau_{\text{MODIS}}$, thereby showing negative biases, particularly in the areas where a dust storm had traveled. This may have arisen due to the cloud pixel screen-out procedure used in retrieving $\tau_{\text{MODIS}}$. Except for the spring episode, the levels of $\tau_{\text{CMAQ}}$ were generally comparable to or slightly higher than those of $\tau_{\text{MODIS}}$. Through US EPA Models-3/CMAQ modeling, the contribution of each particulate chemical species to $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ was investigated. During the summer episode, the high levels of both $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ may have been caused mainly by the high concentrations of (NH$_4$)$_2$SO$_4$ produced over the Chinese urban and industrial centers. In contrast, the high $\tau_{\text{MODIS}}$ and $\tau_{\text{CMAQ}}$ plumes during the winter episode were related to NH$_4$NO$_3$ concentrations over the urban and industrial centers, as well as over Chinese agricultural and livestock farming areas. Finally, $\tau_{\text{MODIS}}$ was greatly correlated with $\tau_{\text{AERONET}}$, indicating the promising potential of the application of the M-BAER algorithm to East Asian air
quality and satellite-based monitoring studies.

In future analysis, given the recent capability for aerosol optical products to be retrieved with very fine resolutions (e.g., 13 × 24 km^2-resolved \( \tau \) from NASA/AURA OMI platform), 3-D photochemistry-aerosol modeling should be conducted with matching fine resolutions, possibly by employing a nested-grid technique (Byun and Ching, 1999). In addition, US EPA has released a ‘spin-off’ version of the CMAQ model, called the CMAQ-MADRID (Model of Aerosol Dynamics, Reaction, Ionization, and Dissolution) model (Zhang et al., 2002, 2004; Pun et al., 2005). The CMAQ-MADRID model still uses the US EPA Models-3/CMAQ frame, but includes alternative aerosol dynamic, microphysical, and chemical processes. As the CMAQ-MADRID model has a more sophisticated SOA formation scheme than the CMAQ model (Zhang et al., 2004; Pun et al., 2002, 2005), the use of the former may produce a different particulate chemical composition and therefore different \( \tau_{\text{CMAQ-MADRID}} \).

In the remote sensing of aerosol optical properties, various types of aerosol optical properties (such as \( \tau \), \( \omega_o \), FMF, and \( \text{reff} \)) are becoming available at higher qualities. For example, FMF was not considered in this study because it has not yet been included in the M-BAER algorithm. However, future studies should consider and attempt to identify the changes in the aerosol size-distribution by using satellite-derived FMF and/or \( \text{reff} \). Furthermore, although satellite remote sensing has a serious limitation in obtaining vertical aerosol extinction profiles, space-borne lidar systems such as CALIPSO have released vertically resolved, spectral aerosol extinction profiles since 2006. The availability of such data will provide further opportunities for atmospheric modelers and the monitoring community to investigate aerosol formation and transport characteristics on both regional and global scales.

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Song, C. H., Maxell-Meier, K., Rodney, R. J., Kapustin, V., and Clarke, A.: Dust composition


Table 1. Parameters of the best-fit Eq. (6) relating $f(RH)$ to relative humidity (RH) in the four seasons.

<table>
<thead>
<tr>
<th>Season</th>
<th>$b_0$</th>
<th>$b_1$</th>
<th>$b_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Spring</td>
<td>-0.01097</td>
<td>0.78095</td>
<td>0.08015</td>
</tr>
<tr>
<td>Summer</td>
<td>0.18614</td>
<td>0.99211</td>
<td>-</td>
</tr>
<tr>
<td>Fall</td>
<td>-0.24812</td>
<td>1.01865</td>
<td>1.01074</td>
</tr>
<tr>
<td>Winter</td>
<td>0.34603</td>
<td>0.81984</td>
<td>-</td>
</tr>
<tr>
<td>Annual</td>
<td>0.33713</td>
<td>0.58601</td>
<td>0.09164</td>
</tr>
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</table>

Table 2. Statistical values among $\tau_{\text{CMAQ}}$, $\tau_{\text{M\&BAER}}$, and $\tau_{\text{NASA}}$ over East Asia.

<table>
<thead>
<tr>
<th></th>
<th>RMSE</th>
<th>MNGE</th>
<th>MB</th>
<th>MNB</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\tau_{\text{CMAQ}}$ vs. $\tau_{\text{M&amp;BAER}}$</td>
<td>$\tau_{\text{M&amp;BAER}}$</td>
<td>$\tau_{\text{NASA}}$</td>
<td>$\tau_{\text{M&amp;BAER}}$</td>
<td>$\tau_{\text{NASA}}$</td>
</tr>
<tr>
<td>Spring</td>
<td>0.27</td>
<td>0.30</td>
<td>39.70</td>
<td>36.05</td>
</tr>
<tr>
<td>Summer</td>
<td>0.26</td>
<td>0.26</td>
<td>65.49</td>
<td>47.56</td>
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<tr>
<td>Fall</td>
<td>0.24</td>
<td>0.26</td>
<td>57.46</td>
<td>70.51</td>
</tr>
<tr>
<td>Winter</td>
<td>0.23</td>
<td>0.23</td>
<td>41.30</td>
<td>40.28</td>
</tr>
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</table>
Table 3. Statistical values among $\tau_{\text{CMAQ}}$, $\tau_{\text{M □ BAER}}$, $\tau_{\text{NASA}}$ and $\tau_{\text{AERONET}}$ at the AERONET sites in East Asia.

<table>
<thead>
<tr>
<th></th>
<th>RMSE</th>
<th>MNGE</th>
<th>MB</th>
<th>MNB</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\tau_{\text{CMAQ}}$</td>
<td>$\tau_{\text{M □ BAER}}$</td>
<td>$\tau_{\text{NASA}}$</td>
<td>$\tau_{\text{AERONET}}$</td>
</tr>
<tr>
<td>Spring</td>
<td>0.25</td>
<td>0.24</td>
<td>0.23</td>
<td>45.51</td>
</tr>
<tr>
<td></td>
<td>0.07</td>
<td>0.10</td>
<td>0.06</td>
<td>41.88</td>
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<tr>
<td></td>
<td>67.95</td>
<td>0.01</td>
<td>0.11</td>
<td></td>
</tr>
<tr>
<td>Summer</td>
<td>0.39</td>
<td>0.65</td>
<td>0.52</td>
<td>82.23</td>
</tr>
<tr>
<td></td>
<td>0.06</td>
<td>0.16</td>
<td>0.18</td>
<td></td>
</tr>
<tr>
<td></td>
<td>45.51</td>
<td>0.01</td>
<td>0.11</td>
<td></td>
</tr>
<tr>
<td>Fall</td>
<td>0.38</td>
<td>0.26</td>
<td>0.26</td>
<td>78.85</td>
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<tr>
<td></td>
<td>0.10</td>
<td>0.20</td>
<td>0.25</td>
<td></td>
</tr>
<tr>
<td></td>
<td>93.55</td>
<td>0.01</td>
<td>0.11</td>
<td></td>
</tr>
<tr>
<td>Winter</td>
<td>0.21</td>
<td>0.36</td>
<td>0.27</td>
<td>49.63</td>
</tr>
<tr>
<td></td>
<td>0.36</td>
<td>0.17</td>
<td>0.17</td>
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</tr>
<tr>
<td></td>
<td>54.31</td>
<td>0.01</td>
<td>0.11</td>
<td></td>
</tr>
</tbody>
</table>

1 indicates that $\tau_{\text{CMAQ}}$ was used as the test variable in Eqs. (10–13).
2 indicates that $\tau_{\text{AERONET}}$ was used as the reference variable in Eqs. (10–13).

**Fig. 1.** Modeling domain in this study.
Fig. 2. Emissions in East Asia: (a) \( \text{SO}_2 \), (b) \( \text{NO}_x \), (c) \( \text{NH}_3 \), (d) NMVOCs, (e) \( \text{BC} \), and (f) \( \text{OC} \).

Fig. 3. MODIS AOD retrieval through the M-BAER algorithm. DEM, SST, and LUT represent the digital elevation model, sea surface temperature, and look-up table, respectively.
**Fig. 4.** Spatial distributions of episode-average $\tau_{\text{M-BAER}}$ (top panels), $\tau_{\text{NASA}}$ (panels in the second row), $\tau_{\text{CMAQ}}$ (panels in the third row), and CMAQ-derived, aerosol column loading (panels in the fourth row) for the four seasons.

**Fig. 5.** Snapshots of the infrared difference dust index (IDDI) with clouds from the NOAA satellite: (a) 7 April, (b) 8 April, and (c) 9 April 2002. The yellow and white colors represent dust plumes and clouds, respectively.
Fig. 6. Spatial and temporal distributions of TOMS AI (top panels) and model-predicted, column mass loading (bottom panels) during a dust storm period (from 7 to 9 April 2002) in East Asia.

Fig. 7. Species-wise, spatial distributions of particulate components for the four seasons. Nitrate and sulfate concentrations in the first and second columns are associated with ammonium (NH$_4^+$) in fine-mode particles. BC and SOAs denote black carbon and secondary organic aerosols, respectively.
Fig. 8. Correlations between $\tau_{\text{MODIS}}$ ($\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$) and $\tau_{\text{CMAQ}}$ for the four seasons: (a) Spring, (b) Summer, (c) Fall, and (d) Winter.

Fig. 9. Probability distribution function (PDF) of $\tau_{\text{MODIS}}$ ($\tau_{\text{M-BAER}}$ and $\tau_{\text{NASA}}$) and $\tau_{\text{CMAQ}}$ in the modeling domain for the four seasons: (a) Spring, (b) Summer, (c) Fall, (d) Winter, and (e) all seasons.
Fig. 10. AERONET sites in East Asia.

Fig. 11.
Fig. 11. Daily variations of $\tau_{AERONET}$, $\tau_{CMAQ}$, $\tau_{M-\text{BAER}}$, and $\tau_{NASA}$ (upper panels) and daily variations of CMAQ-derived particulate composition (bottom panels) in several AERONET sites in East Asia for the four seasons. At the upper panel of each figure, the black bars, solid lines, red circles, and blue squares represent $\tau_{AERONET}$, $\tau_{CMAQ}$, $\tau_{M-\text{BAER}}$, and $\tau_{NASA}$, respectively. In each season, the four sites with the largest number of $\tau_{AERONET}$ were selected.
Fig. 12. Correlations between $\tau_{\text{AERONET}}$ and $\tau_{\text{MODIS}}$ in the four seasons: (a) Spring, (b) Summer, (c) Fall, and (d) Winter.