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# Spatial distributions and seasonal cycles of aerosols in India and China seen in global climate-aerosol model

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# Abstract

A climate-aerosol model is employed to study spatial and temporal variability of aerosol properties over India and China for recent (year 2006) and future conditions (year 2020) under different emission pathways. We present results for aerosol mass concentration

- in different size classes and optical properties for the five different aerosol species treated by the model. Aerosol mass concentration and optical depth have significant contributions from both anthropogenic and natural aerosols. Different species have maxima in different regions, with the highest anthropogenic aerosol concentrations found in Kolkata and elsewhere in the Ganges basin in India and on the northern part
- <sup>10</sup> of the east coast and in the Sichuan basin in China. In India natural aerosols have a maximum in the summer due to higher wind speeds and anthropogenic aerosols have a maximum in the winter due to less efficient wet removal. Surface concentrations are also higher in winter due to the additional reason of lower average boundary layer height. In China seasonal cycles are weaker with natural aerosols having a maximum
- in the spring and sulfate contribution to the aerosol optical depth (AOD) being higher in the latter half of the year. MODIS AOD spatial distributions are reproduced well by the model, except for the Ganges valley with high absorption and for the Thar desert with high dust concentrations. Seasonal cycles compare well qualitatively with MODIS results. The larger AOD in China during the latter half of the year in the year 2006 simulation as compared to the MODIS data can be traced back to sulfate contribution
- with some contribution also from natural aerosols.

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# 1 Introduction

# 1.1 Background

Asia is presently experiencing a high load of aerosols, much of which is of anthropogenic origin. Aerosols have significant effects on human health, the environment and the climate and addressing air quality and climate effects of aerosols simultane-5 ously has become popular in recent years (Ramanathan, 2003; Ramanathan et al., 2007). To quote some numbers from recent studies, outdoor exposure to air pollution is estimated to cause 337 000 premature deaths per year in China and India Ramanathan et al. (2008) and exposure to indoor pollution attributable to solid fuel use is estimated to cause an additional 380 700 deaths in China and 407 100 deaths is India 10 (Ramanathan et al., 2008). Aerosols reduce sunlight coming to the surface in India and China by more than  $10 \text{ W/m}^2$  in most parts of India and China (Ramanathan et al., 2008). Absorbing aerosols heat the atmosphere at elevated levels and increase the absorption of solar radiation when deposited on the glacier surfaces, contributing to glacier melting (Ramanathan et al., 2008), which threatens the water supply and 15 thereby the food security of people living along the rivers receiving melt water flow from the glaciers (Kehrwald et al., 2008). Atmospheric heating caused by absorbing aerosols and surface dimming caused by all aerosols also slow down the hydrological cycle (Chung and Ramanathan, 2006; Xu, 2001). Knowledge of aerosol distributions

<sup>20</sup> at large scale is necessary to evaluate their effects on human health, the environment and the climate.

Aerosols over Asia have been studied in several field campaigns, among others IN-DOEX (Ramanathan et al., 2001), Ace-Asia (Seinfeld et al., 2004) and TRACE-P (Jacob, 2004). Data from several measurement stations are available, including stations
 with long-term high-quality data (Carrico et al., 2003; Ramanathan et al., 2007). Finnish Meteorological Institute (FMI) has performed continuous aerosol measurements at two locations in Northern India: Gual Pahari near New Delhi and the remote Mukteshvar, together with The Energy and Resources Institute of India (TERI) (Hyvärinen et al.,



2009; Komppula et al., 2009; Hyvärinen et al., 2010). University of Helsinki has made urban measurements in India (Mönkkönen et al., 2004, 2005) and Leibniz Institute for Tropospheric Research has made measurements both in India and China (Franke et al., 2003; Müller et al., 2006). There are several studies on aerosol optical properties seen from satellites and AERONET measurement stations (Ramachandran, 2007; Ramachandran and Cherian, 2008). In recent modeling studies, for instance seasonal cycle characterization and climate impact assessment has been made and satellite measurements of aerosol optical depth have been assimilated with chemical transport models (Adhikary et al., 2007; Roeckner et al., 2006b; Adhikary et al., 2008).

#### 10 1.2 Scope of current study

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In this study we use a global aerosol-climate model to investigate the spatial and temporal variability of aerosol properties over India and China, two regions with high aerosol load but somewhat different emission and climate characteristics. Our goal is to provide information about aerosols at large scale, that may be used for air quality and climate research and assessment purposes. We consider both the present-day situation as well as conditions in the future (year 2020) under different emission pathways. The model used is ECHAM5-HAM developed at the Max Planck Institute for

Meteorology (MPI-M) (Stier et al., 2005).

We present results for surface aerosol concentrations in the form of PM<sub>1.0</sub>, PM<sub>2.5</sub> and
 PM<sub>10</sub> and the contributions to these of the chemical components treated by our model: sulfate (SU), black carbon (BC), organic carbon (OC), mineral dust (DU) and sea salt (SS). We will also study optical properties: aerosol optical depth (AOD) absorption aerosol optical depth and Angstrom parameter. AOD will be split up according to the chemical components. In addition to a simulation of present-day Asia (year 2006), we

present results for a simulation without any anthropogenic emissions and for simulations with two future scenarios for the year 2020: the reference scenario created by the authors of the Regional Asian Emission Inventory (Ohara, 2007; GEIA/ACCENT 2008) and a scenario with anthropogenic emissions of all aerosol species decreasing by 2



per cent a year. The modeled AOD is compared to the AODs obtained from MODIS Terra. We also make qualitative comparisons to some ground-based measurements.

We aim to present a wide-perspective look at these aerosol properties, which affect and depend on human life and activities, weather and climate on an essentially regional

- <sup>5</sup> scale. Local effects are naturally outside the scope of this study. While we are presenting new scientific results we also hope to highlight and clarify some features that are harder to see in more detailed studies. We use a well-established model, a recent emission inventory and scenarios based on the most up-to-date information available on the fast-changing socioeconomic situation in these countries. Earlier large-scale
- <sup>10</sup> model were presented by Adhikary et al. (2007) and Carmichael et al. (2009). The current study uses a detailed aerosol model and a recent emission inventory and studies seasonal cycles as well. Our model also accounts for the effect of aerosols on meteorological conditions.

The model and the simulations are presented in Sect. 2. The Regional Asian Emission Inventory and the emission scenarios used are presented in Sect. 3. Results for present-day Asia are presented in Sect. 4 and for the future simulations in Sect. 5. The main findings are summarized in Sect. 6.

# 2 The model and simulations

The ECHAM5-HAM model is an atmospheric general circulation model (GCM) cou-<sup>20</sup> pled with an aerosol model. It predicts the evolution of an ensemble of microphysically interacting internally- and externally-mixed aerosol populations as well as their sizedistribution and composition. ECHAM5-HAM is described by Stier et al. (2005) and the atmospheric part ECHAM5 in Roeckner et al. (2006a) and Roeckner et al. (2003). The model treats five aerosol species: sulfate, black carbon, organic carbon, mineral <sup>25</sup> dust and sea salt The first three mentioned components are emitted by both anthro-

<sup>25</sup> dust and sea salt The first three mentioned components are emitted by both anthropogenic and natural sources and the two last mentioned by wind-induced formation from natural sources only. The size-distribution is represented by a superposition of



seven log-normal modes (Vignati et al., 2004). Emissions of all species except sulfur compounds are in the form of particulate matter. Except for marine DMS emissions, 97.5% of sulfur emissions are in the form of  $SO_2$  and 2.5% in the form of primary sulfate. All emissions are distributed evenly inside the model grid box. The aerosol

<sup>5</sup> effects on radiation are only considered for the shortwave part of the spectrum. The atmospheric part is a fifth-generation climate model that has evolved from the model of the European Centre for Medium-Range Weather Forecasts (ECMWF).

Stier et al. (2005) report that the model generally agrees well with observations of the global aerosol system. For example, the global annual mean AOD 0.14 is in excellent agreement with AERONET measurements (0.14) and a composite of MODIS-MISR

- <sup>10</sup> agreement with AERONET measurements (0.14) and a composite of MODIS-MISR satellite retrievals (0.16). The model is also able to reproduce the main regional features, although non-negligible deviations exist. In Sect. 4 of this article some new comparisons between simulation results and observations are provided.
- In this study we present results from four simulations with different aerosol emissions, two for the year 2006 and two for the year 2020. The model is run at T42L19 resolution, meaning a horizontal grid spacing of ~2.8° and 19 vertical levels in a hybrid sigma/pressure coordinate system with the top level at 10 hPa. For greenhouse gas concentrations we use the IPCC SRES scenario A1B (Nakicenovic et al., 2000). The sea-surface temperatures and sea ice fields are taken from a simulation with the coupled ECHAM5-MPI-OM (Max Planck Institute Ocean Model) atmosphere-ocean GCM (Marsland et al., 2003). All simulations were run for five years.

#### 3 Emission inventory and scenarios

The REAS emission inventory includes inventories for anthropogenic emissions for  $SO_2$ , black carbon and organic carbon treated by our model and additionally for  $NO_x$ ,

<sup>25</sup> CO and NMVOC (nonmethane volatile organic compounds) from fuel combustion and industrial sources. REAS has been produced with a methodology using emission factors, activity data and removal efficiency of emission controls. Emissions are given for



1980–2003 and for three future scenarios reaching 2020: a reference scenario (REF), a policy success case (PSC) and a policy failure case (PFC). REAS is the first inventory presenting historic, present and future emissions of multiple aerosol species in Asia on the basis of a constant methodology. Another widely used inventory for Asia is that by Streets et al. (2003). See Ohara et al. (2007) for a review on other inventories.

 Streets et al. (2003). See Ohara et al. (2007) for a review on other inventories. Seasonal variations in anthropogenic emissions are not considered in the REAS inventory. Thus, for example emissions due to increased heating in colder regions during winter or emissions due to increased tourism in the summer are omitted. In spite of this limitation, the seasonal variations in aerosol properties in our simulations
 seem quite reasonable (Sect. 4.1). Variations within the year shown in the following plots are purely due to meteorological conditions influencing aerosol processes and variations in natural aerosol emissions.

Anthropogenic emissions of SO<sub>2</sub>, black carbon and organic carbon are overall very large for China and India in the REAS inventory both for past years and the future scenarios. In the year 2000 SO<sub>2</sub> emissions were 27.6 Mt for China and 6.1 Mt for India.

- BC emissions were 1.1 Mt for China and 0.8 Mt for India while organic carbon emissions were 2.6 Mt for China and 3.3 Mt for India. The main source for the strikingly high  $SO_2$  emissions in China was coal combustion in power plants and industrial facilities, most of which lacked desulfurisation systems, something that has improved since the
- REAS inventory was made (Lu et al., 2010). In India coal combustion in power plants and oil combustion in industry gave major contributions to the total SO<sub>2</sub> emissions. Taking Japan as a reference, the mean emission factor for SO<sub>2</sub> emissions from coal combustion in power plants in 2000 was 833.6 g/GJ in China and 550.9 g/GJ in India compared to 26.2 g/GJ in Japan and in industry 934.2 g/GJ for China and 250.1 g/GJ
- for Japan. The mean emission factor for SO<sub>2</sub> emissions from oil combustion in industry was 148.7 g/GJ in China and 562.1 g/GJ in India compared to 35.9 g/GJ in Japan. The major source for the very large BC and OC emissions was domestic fuel combustion for both countries, where in China both biomass and coal accounted for large shares and in India biomass burning was dominant. Domestic fuel combustion when cooking



with fuels such as wood, crop residues, dung and coal is often done in stoves with very incomplete combustion and poor ventilation with the airborne pollution released to the living area and the atmosphere. Also fuel efficiency is often very low. Even stoves using wood, which is one of the cleaner biofuels, typically produce carbon monoxide and hydrocarbon emissions 50 times higher compared to cooking an equivalent meal

and hydrocarbon emissions 50 times higher compared to cooking an equivalent meal on a gas stove, as an example (Bruce et al., 2002). In urban areas also traffic is a heavy emitter, especially of the NO<sub>x</sub> compounds not treated by our model.

Large emissions imply that there is a lot of potential for reductions. Most  $SO_2$  emissions could be eliminated by introducing desulfurizations systems. The above compar-

- ison of emission factors with Japan indicates that remarkable emission reductions can be made even in the current fossil fuel based facilities. Furthermore, in India and China there is enormous potential for renewable energy such as wind and solar power, which help to curb both the air quality and climate effects of electricity production. Household emissions of BC and OC could be reduced by replacing solid fuels with cleaner energy sources (e.g. Project Surya, Edwards et al., 2007) and by using stoves of improved
- design (Ramanathan and Feng, 2009; Hanbar and Karve, 2002).

In view of international treaties on climate and air quality, the emissions will likely be reduced, following e.g. China's outspoken goal on sulfate emission reductions (Xin and Stone, 2008). To study effects of emission reductions we developed for this study a scenario with anthropogenic emissions of sulfur, BC and OC all decreasing by 2 per

- <sup>20</sup> scenario with anthropogenic emissions of sulfur, BC and OC all decreasing by 2 per cent per year in Asia throughout. We call this scenario the –2% per year scenario. A lot faster emission reductions are also possible. Effects of large emission cuts are already considered in some studies, e.g. in Jacob (2004); Ramanathan and Carmichael (2008).
- We present results from four different simulations: simulations of the years 2005– 2009 with (1) REAS prediction inventory emissions for the year 2006 and (2) completely without anthropogenic emissions; and simulations of the years 2019–2023 with (3) the REAS REF scenario and with (4) the -2% per year scenario. Anthropogenic emissions outside Asia are set to zero, which causes some negative bias to aerosol concentrations. However, this bias is likely to be very small as anthropogenic emissions are still



treated for a much larger region than the one analysed (Ohara et al., 2007). The first year of each simulation is used as spin-up and results are presented for the remaining four years.

# 4 Aerosol characteristics in present-day India and China

In the following we present characteristics of aerosol concentrations and optical properties in present-day Asia. The domain-mean results we report for India are averages over the longitude-latitude box stretching from 5° N to 35° N and from 65° E to 90° E. The results for China are averages over the two longitude-latitude boxes 20–45° N × 95–125° E and 35–45° N × 75–95° E. For plotting convenience the Chinese spatial distribution tions are shown only from 95° E eastward.

We present results for mass concentration in different size classes  $PM_{1.0}$ ,  $PM_{2.5}$  and  $PM_{10}$  (Figs. 2, 4, 8 and 9) and the optical properties: AOD (Figs. 3, 5 and 10), absorption AOD (Abs) and Angstrom parameter (Ang) (Fig. 11). All AOD values presented here are values at  $\lambda = 550$  nm. We split up the concentrations in different size classes and AOD to the five different aerosol species involved. As only a negligible part of sulfate, black carbon and organic carbon mass concentration can be found in sizes  $D > 1 \mu m$ , we only plot the seasonal cycles of these aerosol species for a single size class,  $PM_{10}$ . Spatial mass concentration distributions are shown for  $PM_{2.5}$ , which was chosen because in this size class the maximum concentrations of anthropogenic and natural aerosols were closest to each other.

It is not self-evident how to distribute the AOD between the aerosol species. A simple diagnostic of splitting up the AOD according to volume fractions in the internally mixed particles is used. We split up the AOD of aerosol associated water to the hygroscopic species sulfate and sea salt based on concentrations and relative humidity according to

the formula in Jacobson et al. (1996). This simplified scheme is a choice among many possibilities, and though it underestimates the contribution of black carbon to AOD as the internally mixed assumption enhances absorption, its advantage lies in being



simple and in our opinion, accurate enough for the other species. A better estimate of absorption by black carbon can then be obtained from the absorption AOD.

Simulated AOD is compared to measurement results from MODIS and AERONET analysed in the scientific literature. Most of these results are in form of averages over

- some years. We also compare our results to a modeling study by Adhikary et al. (2007), which considers the time frame from September 2004 to August 2005. The different time frames obviously create some discrepancy, but for the climatological aerosol properties studied here, the discrepancy is most probably small. In Ramachandran (2007) it can be seen that the standard deviation of AOD monthly averages from the 5-year
- mean taken between 2001–2005 is much smaller than the average value of AOD for all 20 locations in India considered by the study. We also perform new comparisons of average AOD spatial distribution in India and China with data from the newer MODIS C5 dataset. The MODIS AODs presented consist of collection 4 data for years 2001– 2005 over India (obtained from Ramachandran, 2007 and Ramachandran and Cherian,
- 2008), for 2000–2005 over China (obtained from Kim et al., 2007), and collection 5 15 data for year 2006 over India and China (obtained from GIOVANNI online data system, http://disc.sci.gsfc.nasa.gov/giovanni). The resolution of the MODIS data is  $1^{\circ} \times 1^{\circ}$ .

#### Spatial distributions 4.1

In India the largest PM<sub>2.5</sub> and AOD values occur in the Ganges basin and in the Kolkata region, with the annual grid box average  $PM_{2.5}$  value reaching 60  $\mu$ g/m<sup>3</sup> and AOD 20 reaching values over 0.4 (Figs. 2 and 3). Sulfate, black carbon and organic carbon concentrations all peak in the same region, decreasing when going north or towards the sea. The largest mean sulfate, black carbon and organic carbon concentrations at this resolution are over  $8 \mu g/m^3$ ,  $14 \mu g/m^3$  and  $45 \mu g/m^3$ , respectively (Fig. 2). Of the species emitted by anthropogenic sources, sulfate is distributed more widely than the carbonaceous species. Sulfate concentrations at sea far off the coast are often half of the concentrations in regions close to the sources while concentrations of black and organic carbon drop to a much smaller fraction of that in the source regions just outside

Discussion Paper 11, 4017-4057, 2011 Asian aerosols S. V. Henriksson et al. **Discussion** Paper **Title Page** Abstract Introduction Conclusions References **Figures** Tables **Discussion** Paper Back Full Screen / Esc **Discussion** Paper **Printer-friendly Version** Interactive Discussion

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the coast. The main reason for this difference is that most sulfur emissions are in the form of  $SO_2$  and therefore have a chance to get transported higher and longer than the carbon aerosols emitted as primary matter. It is well known that aerosol lifetime in the atmosphere is strongly dependent on emission height (see e.g. Balkanski, 1991). For this reason secondary organic aerosols (SOA) would probably also spread wider than primary particle emissions, if the SOA were not formed at the emission site as in the model. Dust concentrations peak in the Thar desert region in the west and in the north on the Tibetian plateau. Sea salt concentrations are highest above the sea, and higher over the Arabian sea than over the Bay of Bengal due to higher winds over the Arabian

10 sea.

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Our sulfate concentrations are similar to those in the modeling study by Adhikary et al. (2007), but black and organic carbon concentrations are larger. At least part of the explanation of this difference lies in the different emission inventories. The REAS inventory (Ohara et al., 2007) used in this study has black and organic carbon emissions over 50% bigher than the inventory organed by Streets et al. used in Adhikary et al.

- over 50% higher than the inventory created by Streets et al. used in Adhikary et al. (2007). Above, we noted that in our simulations, the mass concentration of black and organic carbon decreases faster with increasing distance from source regions than the mass concentration of sulfate. In this respect, it is of interest to consider observational results for black and organic carbon. INDOEX results for Kaashidoo Climate Obser-
- vatory (KCO, 5° N, 73.5° E) in the Maldives southwest of the Indian peninsula showed PM<sub>1</sub> mass consisted of 32% sulfate, 26% organic carbon, 14% black carbon, 10% mineral dust, 1% sea salt and 13% ammonium and fly ash on average in February 1999 (Ramanathan et al., 2001). In February the winter monsoon winds blow from northeast transporting aerosols from India to KCO. In Moorthy et al. (2004) it was reported that
- <sup>25</sup> during the intermonsoon season in March–April, when winds are weaker and therefore less capable to transport aerosols from sources on land, the black carbon mass fraction was as low as 2.2% southwest of India in the Arabian Sea compared to 7–10% at the nearby coastal location of Trivandrum. Also the measurements at the background station in Mukteshvar measured by FMI report 2–4% mean mass fractions of black



carbon in  $PM_{2.5}$  (Hyvärinen et al., 2009). Beegum et al. (2009) reported monthly mean black carbon concentrations measured at eight stations for January 2006 to May 2006 and found large variations within the country and between different months, with an average concentration of 0.065 µg/m<sup>3</sup> over the Arabian sea in May and 27 µg/m<sup>3</sup> in Delhi 5 in January. Considering the results cited in this paragraph we find the differences in

our spatial patterns of sulfate and carbonaceous aerosols qualitatively consistent with field measurements, which generally indicate smaller mass fractions of black carbon in remote regions.

Ramachandran and Cherian in their study on MODIS Terra AOD data (coll. 4) for 2001–2005 in India (Ramachandran, 2007) find the smallest annual mean AOD of 0.28 in northeast India followed by 0.35 in southern India. The largest annual mean AOD 0.49 is found in eastern India followed by 0.47 in northern India. Our simulation results for the years 2006–2009 have the same qualitative features (Fig. 3), though the values are somewhat lower in the simulation. It should be noted, however, that at least over

- some urban locations in India the MODIS AODs (coll. 4) at 550 nm tend to be overestimated when compared to ground based AERONET measurements (Ramachandran, 2007). Over India similar AOD pattern for 2006 can be seen in the MODIS coll. 5 data (Fig. 6), though AOD values in the Ganges basin are higher, 0.5–0.7 and values in the Thar desert area are a lot higher than in the simulation. According to Levy et al. (2010)
- it is expected that the MODIS coll. 5 AODs are underestimated over semi-arid areas with high dust concentrations, hence the model underestimate in this region might be even larger than suggested by Fig. 6.

In China the largest  $PM_{2.5}$  mean values are found close to the east coast (Fig. 4). Yearly average  $PM_{2.5}$  values around Beijing exceed 55 µg/m<sup>3</sup>. Sulfate, black carbon

<sup>25</sup> and organic carbon concentrations exceed 35  $\mu$ g/m<sup>3</sup>, 6.5  $\mu$ g/m<sup>3</sup> and 20  $\mu$ g/m<sup>3</sup>, respectively. The Sichuan basin located around 105° E, 30° N also has a local PM<sub>2.5</sub> maximum and the largest local annual mean sulfate concentration and AOD can be found there (Fig. 5). The large positive difference between mean AOD in the Sichuan basin and the east coast is explained by higher relative humidity and hygroscopic growth of aerosols



due to water condensing on sulfate. As in India, sulfate is more widely spread than black and organic carbon. Dust concentrations peak in the north near or outside the Mongolian border and sea salt concentrations peak over the sea.

Meng et al. in Meng et al. (2007) review aerosol concentrations from other measure-<sup>5</sup> ments made in Beijing, Xi'an, Hong Kong and Koashung with measurements being done between 1998 and 2006. The average black carbon concentrations in these urban areas were 4.0–9.7  $\mu$ g/m<sup>3</sup> and the organic carbon concentrations 10.4–61.9  $\mu$ g/m<sup>3</sup>, higher or roughly the same as our simulated grid box averages (Fig. 4).

In their study on MODIS data averaged from April 2000 to June 2005, Kim et al. (2007) also find a local maximum at the Sichuan basin. The observed average AOD 0.7–0.8 is smaller than in our simulations (maximum AOD = 1.0, see Fig. 5) but larger than that for most part of the east coast. Figure 7 shows the MODIS coll. 5 AODs over China in 2006. Here the AOD reaches the value 1.0 in the Sichuan basin. Our simulated AOD spatial pattern is qualitatively consistent with this MODIS-derived data,

mean AOD being larger on the northern part of the east coast and on the south coast compared to the south-eastern coast. Values west of the Sichuan basin are very low both in the simulation and MODIS data. Our results have less local features, which is not surprising, since the MODIS data have a resolution of 1° × 1°. PM<sub>2.5</sub> values in the Sichuan basin are lower than those on the northern east coast and the higher AOD is
 a result of hygroscopic growth of sulfate aerosols.

#### 4.2 Seasonal variations

The annual cycle of  $PM_{2.5}$  and AOD in India consists of superimposed seasonal cycles of fine anthropogenic particles and coarse natural particles, with the seasonal cycle for anthropogenic aerosols (most of sulfate, BC and OC; see Sect. 5) having its maximum <sup>25</sup> in the winter and that of natural aerosols having its maximum in the summer (Figs. 8– 10). The overwhelming majority of sulfate, BC and OC particles lie in the *D* < 1 µm regime and therefore we plot them only in one size class  $PM_{10}$ . Mineral dust and sea salt distributions contain a lot of larger particles and therefore we plot also their contributions in the  $PM_1$  and  $PM_{2.5}$  classes separately (Fig. 9). From the PM figures



it can be seen that the mass of natural aerosols mineral dust and sea salt lies mainly in the  $D > 1 \,\mu\text{m}$  regime. The seasonal cycles in India are strong, with monthly average concentrations of sulfate, black carbon and organic carbon varying by more than a factor of 2 and of mineral dust and sea salt by approximately a factor of 10 within the

- year. The strong maximum of anthropogenic aerosols in the winter is attributed mainly to lack of rain and wet removal of aerosols and the strong maximum of natural aerosols in the summer is mainly attributed to high wind speed (see Fig. 12). Another important factor at work is the height of the boundary layer, which is lower in the winter and thereby causes higher surface concentrations. That the total columnar anthropogenic
- <sup>10</sup> aerosol concentrations are indeed higher in the winter can be seen for instance in the AOD values for the individual species (Fig. 10), which have similar seasonal cycles. The sulfate AOD seasonal cycle is weaker indicating that removal of aerosols in the regions with high sulfate concentrations does not vary as much as elsewhere. The seasonal cycle can also be clearly seen in the absorption AOD monthly mean values,
- <sup>15</sup> which vary from 0.015 in the summer to 0.037 in the winter, and in the Angstrom parameter monthly mean values, which vary from 0.8 in the summer to 1.6 in the winter (Fig. 11). The shape of the Indian average AOD curve resembles well the majority of MODIS-derived curves for separate locations in Ramachandran (2007), though our winter maximum seems to be relatively higher in comparison to the summer maximum.
- <sup>20</sup> This might be due to the emissions in the year 2006 being higher than in years just before as Ramachandran (2007) studies the period between 2001 and 2005.

In their modeling study on southern Asia for the period September 2004 to August 2005 Adhikary et al. (2007) obtain maxima for the anthropogenic aerosols species for most locations in October–December, whereas our maximum is perhaps shifted

forward by one month to November–January. The average we obtain for India is consistent with this. As to natural aerosols, Adhikary et al. find a maximum in the summer, which is consistent with our results. AOD seasonal cycles from MODIS satellite measurements (Ramachandran, 2007) and AERONET ground-based measurements (Singh et al., 2004) show summertime maxima mostly between May and August and



secondary maxima in the winter between November and January. Our results are consistent with these measurements.

In China, the seasonal cycles are not quite as strong as in India (Figs. 8–11). Black carbon concentrations are higher in winter than in summer. The organic carbon con-<sup>5</sup> centration average is not changing as much and has a few smaller maxima. The average sulfate concentration is high in October–January. AOD has maxima in February– May and in August–October, with the first maximum explained more by natural aerosols and the second maximum explained more by sulfate.

Comparing the AOD seasonal cycle to the MODIS-derived result for data from April 2000 to June 2005 in Kim et al. (2007), the shapes of the AOD curves match well in the first half of the year. In Kim et al. (2007) the spring maximum occurs during the same months as in our simulations and is just below 0.45, higher than in our simulation, then decreases to the level 0.25–0.3 like in the MODIS results, rising again to levels 0.35– 0.4 for July–August (Fig. 10). The remaining difference is partly caused by different averaging domains. There is discrepancy in the second half of the year, which does

- not have to mean that either one is wrong as the averaging domain in Kim et al. (2007) is located further north and east than in our study. The high AOD in the latter half of the year might be due to the fact that Chinese  $SO_2$  emissions were higher in 2006 than during the period from April 2000 to June 2005. The large effect of the magnitude
- of Chinese SO<sub>2</sub> emissions on AOD in the latter half of the year as well as emission trends are discussed in the article by Lu et al. (2010). This might partly explain the difference between simulation and MODIS results for August and September seen in Fig. 7. Another possible partial explanation of high AODs in the latter half of the year is the recently found fact that ECHAM5-HAM overestimates SO<sub>2</sub> concentrations in some
- regions (Kazil et al., 2010). In the next section it will be shown that reducing sulfate emissions affect the AOD more in the latter half of the year.

Comparison of simulated concentrations over India and China shows that sulfate concentrations are far larger in China than India with monthly means varying between 7  $\mu$ g/m<sup>3</sup> and 11  $\mu$ g/m<sup>3</sup> for China and between values less than 2  $\mu$ g/m<sup>3</sup> and just over



5 μg/m<sup>3</sup> for India. The black and organic carbon mean concentration is higher in India than in China throughout the year and the difference is much bigger in winter than in summer. In January the mean black carbon concentration is 4.6 μg/m<sup>3</sup> in India and 2.2 μg/m<sup>3</sup> in China and in July 1.5 μg/m<sup>3</sup> in India and 1.2 μg/m<sup>3</sup> in China. The largest dust concentrations are seen in March–May for China and in May–July in India, with highest monthly mean dust concentration in the  $D < 10 \,\mu$ m size class being 16 μg/m<sup>3</sup> in June for India and 19 μg/m<sup>3</sup> in April for China. Sea salt concentrations are larger in India but the values are not comparable, because a larger fraction of the modelled Indian domain is covered with ocean than in the Chinese case.

The four-year monthly mean PM<sub>2.5</sub> in India has on average relative contributions of 15% sulfate, 13% black carbon, 43% organic carbon, 17% mineral dust and 12% sea salt. In China the relative contributions are 42% sulfate, 8% black carbon, 25% organic carbon, 22% mineral dust and 3% sea salt.

We wish to point out a few things that need to be considered when interpreting the results. Firstly, although the natural aerosols mineral dust and sea salt contribute significantly to the PM and AOD mean values, most of their contribution comes from regions with very low population density and low agricultural activity. Therefore the anthropogenic contribution to harmful pollution for human beings is larger than national averages would suggest. Secondly, most anthropogenic particles are in the size class

- <sup>20</sup> PM<sub>1</sub>, but for the natural particles produced by action of the wind, less than 10% of the concentration comes from particles with  $D < 1 \,\mu\text{m}$  and over 50% comes from particles with  $D > 2.5 \,\mu\text{m}$ . This suggest that PM<sub>1</sub> and either PM<sub>10</sub> or PM<sub>10-1</sub> (i.e. concentration of particles with  $1 \,\mu\text{m} < D < 10 \,\mu\text{m}$ ) might be more optimal metrics of aerosol concentrations than PM<sub>2.5</sub>, a point that has been made also from a health impact perspective
- 25 (Perez et al., 2009). Thirdly, the mean AOD is calculated for making comparisons between results from different sources, not for making deductions on radiative properties. The transmissivity of aerosols depend non-linearly on AOD and radiative forcing yet on the strenght of incoming solar radiation strength, relation between scattering and absorption and asymmetry in scattering.



#### 5 Results for future scenarios and without anthropogenic emissions

In this section we will present results from a simulation with the REAS reference scenario for 2020, from a simulation using the –2% per year scenario for 2020 (with results for these two simulations being presented for 2020–2023) and from a simulation with no anthropogenic emissions for the years 2006–2009. We compare these results to the present-day simulation results presented in the previous section. In the REAS reference scenario, emissions of SO<sub>2</sub>, black carbon and organic carbon continue to increase in India, year 2020 emissions being 43.4%, 8.5% and 8.4% higher, respectively, than those in year 2006. However, the Chinese emissions in year 2020 are smaller than those in the year 2006, by 22.3% for SO<sub>2</sub>, 38.8% for black carbon and 39.4% for organic carbon. In the –2% per year scenario anthropogenic sulfate, black carbon and organic carbon emissions decrease with a rate of 2 per cent per year from 2006 to 2020 yielding 24.6 per cent reductions in total. We present the results for PM<sub>2.5</sub> and AOD in Fig. 13.

Most of the sulfate, black carbon and organic carbon aerosols are of anthropogenic origin. In the simulation with no anthropogenic emissions sulfate monthly mean concentrations are less than 0.5 μg/m<sup>3</sup> for both India and China, organic carbon monthly mean concentrations are of the order 1 μg/m<sup>3</sup> or less and black carbon monthly mean concentrations are less than 0.1 μg/m<sup>3</sup> (see Tables 1 and 2). Monthly average PM<sub>2.5</sub>
 values are still around 3–15 μg/m<sup>3</sup> and AOD values around 0.04–0.18 due to natural aerosols.

Relative changes of  $PM_{2.5}$  and AOD in the differents scenarios are shown in Tables 1 and 2, separated by aerosol species. As expected, concentrations of an aerosol species and its AOD generally decrease when its emissions are decreased. In India,

<sup>25</sup> monthly sulfate concentrations are higher for all months in the REAS scenario. For black and organic carbon the nearly 10% increases compared to year 2006 cause the concentrations to rise mainly in October–February while concentrations during the rest of the year are less affected (Fig. 12, concentrations of different species not shown).



Anthropogenic AOD increases. In China sulfate, black carbon and organic carbon mean concentrations decrease for all months in the REAS scenario compared to the present-day simulation (Fig. 12, concentrations of different species not shown). The AODs of the anthropogenic species also decrease.

- In the -2% per year scenario sulfate, black carbon and organic carbon monthly mean concentrations are all lower than in the present-day simulation except for the February mean sulfate concentration in China. The four-year mean PM<sub>2.5</sub> value decreases by 17.6% in India and 20.6% in China. Mean sulfate, black carbon and organic carbon concentrations decrease by 22.5%, 23.0% and 21.2% in India and 25.8%, 25.8% and
- 10 22.5% in China, respectively. Mineral dust and sea salt concentrations vary less, as expected. Thus, in both India and China the sulfate, black carbon and organic carbon concentrations experience a relative decrease approximately the same as the corresponding relative decrease in emissions. Changes in AOD for the different aerosol species are slightly but not much different than changes in concentrations. Total AOD
- is affected both by changes in anthropogenic and natural emissions, with the latter depending on meteorological conditions, and therefore the total AOD does not necessarily decrease by the same amount as anthropogenic emissions. In this case the mean AOD decreases by 21.7% in India and by 22.0% in China.

We can compare the AOD seasonal cycles of China and India in these future sim-

- <sup>20</sup> ulations (Fig. 13) to the MODIS-derived seasonal cycles (Ramachandran, 2007; Kim et al., 2007). The AOD curve for India in the –2% per year simulation resembles, like in the case with REAS 2006 emissions, in shape the MODIS-derived curves for most locations studied in Ramachandran (2007), which could mean that the real emissions in 2001–2005 have been somewhere between REAS inventory emissions for 2006 and
- the -2% per year scenario emissions. The AOD curve for China resembles MODIS results for a different averaging domain (Kim et al., 2007), which could be due to different domains and differences in SO<sub>2</sub> emissions balancing out each other (Lu et al., 2010).



# 6 Conclusions

We have presented results for aerosol seasonal cycles and spatial distributions in India and China seen in a global aerosol-climate model using the Regional Asian Emission Inventory. The modelled aerosol species were sulfate, black carbon, organic carbon, mineral dust and sea salt. Sulfate, black carbon and organic carbon particles were 5 found mostly in the  $D < 1 \, \mu m$  regime, whereas the natural particles typically had diameters larger than 1 µm and a large part of their mass in size classes larger than PM<sub>2.5</sub>. The Indian and Chinese aerosol characteristics and seasonal cycles differed from each other. India had higher concentrations of black and organic carbon and China had higher concentrations of sulfate. The Indian aerosol seasonal cycle con-10 sisted of a strong natural cycle with maximum in the spring-summer (May-July) and a strong anthropogenic cycle having a maximum in the winter (November-January). The seasonal cycle of natural aerosols was guite strong in China too with a maximum in spring (March-May) but the seasonal cycle of anthropogenic aerosols was not as strong with area-mean quantities varying less from month to month.

An interesting feature in the simulations is the difference between spatial patterns of sulfate and carbonaceous aerosols, sulfate concentrations remaining high far away from sources and black and organic carbon concentrations decreasing faster with distance. We attribute this to the fact that SO<sub>2</sub> can be transported higher and further away from the sources before being converted to particles compared to carbonaceous aerosols that are emitted in particulate form. Another interesting feature was the high AOD in the Sichuan basin in China caused mainly by high sulfate concentrations combined with high relative humidity.

Our simulations reproduce MODIS-derived AOD seasonal cycles published in literature reasonably well for India and for a longitude-latitude domain in China, when using the -2% per year scenario. In August and September the discrepancy observed for the present-day Chinese case is probably due to larger SO<sub>2</sub> emissions in the simulation than in the real situation observed. Of course the absolute value of AOD should not be



used as a single metric for air quality since the same AOD can be caused by a number of different aerosol distributions. In India, monthly average AOD is largest when fine anthropogenic particle pollution is at its lowest.

We performed future simulations for 2020 using two scenarios: the REAS reference scenario and a -2% per year scenario. In general, concentrations in the future simulations had the same trends as emissions. Cuts in anthropogenic emissions also influenced the seasonal cycles, mainly because of different relative contributions of anthropogenic and natural aerosols at different times of the year. In India mostly wintertime concentrations and AOD were effected and in China emission cuts affected the AOD most during the latter half of the year. However, emission cuts led to lower PM<sub>2.5</sub>

and AOD during the latter half of the year. However, emission cuts led to lower PM and AOD during the entire year in both countries.

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In summary, we have presented an analysis illustrating the contributions of anthropogenic and natural aerosols to different aerosol size classes and aerosol optical depth in large scale in India and China and how changes in anthropogenic emissions influence the concentrations, AOD and their seasonal cycles.

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**Table 1.** Percentage changes of mass concentration and AOD of different aerosol species in

 India. Change from simulation of present-day Asia in the different scenarios.

	REAS REF	–2% per year	anthr. emiss. = 0
PM2.5	+11.6	-17.6	-67.1
S0 <sub>4</sub> 2.5	+43.2	-22.5	-87.1
BC2.5	+13.7	-23.0	-99.0
OC2.5	+12.5	-21.2	-90.8
DU2.5	-5.6	-6.4	-4.0
SS2.5	-11.6	-7.5	-6.8
AOD	+13.6	-21.7	-65.2
AOD_SO4	+31.0	-26.6	-83.2
AOD_BC	+7.0	-27.3	-97.8
AOD_OC	+5.0	-25.3	-87.5
AOD_DU	-17.3	-14.9	-2.0
AOD_SS	-5.6	-10.2	-29.3

Discussion Paper	AC 11, 4017–4 Asian a	ACPD 11, 4017–4057, 2011 Asian aerosols		
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**Table 2.** Percentage changes of mass concentration and AOD of different aerosol species in

 China. Change from simulation of present-day Asia in the different scenarios.

	REAS REF	–2% per year	anthr. emiss. = 0
PM2.5	-16.2	-20.6	-69.2
S0 <sub>4</sub> 2.5	-20.6	-25.8	-96.7
BC2.5	-33.9	-25.8	-98.3
OC2.5	-29.1	-22.5	-85.2
DU2.5	+9.2	-10.4	+0.1
SS2.5	-5.6	+1.3	+6.0
AOD	-12.3	-22.0	-81.0
AOD_SO4	-14.7	-24.7	-93.3
AOD_BC	-16.0	-25.7	-94.7
AOD_OC	-12.4	-20.9	-76.6
AOD_DU	+7.2	-11.7	+2.8
AOD_SS	-5.7	-6.6	-44.7

![](_page_28_Figure_0.jpeg)

**Fig. 1.** Map of South and East Asia including India and China. Results presented for India are averages over the area  $5-35^{\circ}$  N ×  $65-90^{\circ}$  E and results for China are averages over the areas  $20-45^{\circ}$  N ×  $95-125^{\circ}$  E and  $35-45^{\circ}$  N ×  $75-95^{\circ}$  E.

![](_page_28_Figure_2.jpeg)

![](_page_29_Figure_0.jpeg)

![](_page_29_Figure_1.jpeg)

Fig. 2. Total surface  $PM_{2.5}$  in India and contributions to it from different aerosol species (fouryear average, in units  $\mu g/m^3$ ).

4046

![](_page_30_Figure_0.jpeg)

![](_page_30_Figure_1.jpeg)

Fig. 3. Total AOD in India and contributions to it from different aerosol species (four-year average, unitless).

![](_page_31_Figure_0.jpeg)

![](_page_31_Figure_1.jpeg)

**Fig. 4.** Total surface  $PM_{2.5}$  in China and contributions to it from different aerosol species (fouryear average, in units  $\mu g/m^3$ ).

4048

![](_page_32_Figure_0.jpeg)

![](_page_32_Figure_1.jpeg)

Fig. 5. Total AOD in China and contributions to it from different aerosol species (four-year average, unitless).

![](_page_33_Figure_0.jpeg)

Fig. 6. AOD in India according to MODIS C5 data for 2006 (annual average, unitless).

![](_page_33_Figure_2.jpeg)

![](_page_34_Figure_0.jpeg)

Fig. 7. AOD in China according to MODIS C5 data for 2006 (annual average, unitless).

![](_page_34_Figure_2.jpeg)

![](_page_35_Figure_0.jpeg)

![](_page_35_Figure_1.jpeg)

**Fig. 8.** Monthly mean surface  $PM_{10}$  in units  $\mu g/m^3$  and contributions to it from different aerosol species for China (dashed line) and India (solid line), averaged over four years.

![](_page_36_Figure_0.jpeg)

**Fig. 9.** Contributions of mineral dust and sea salt to the four-year mean surface mass concentration in size classes  $PM_{2.5}$  and  $PM_{1.0}$  for China (dashed line) and India (solid line), in units  $\mu g/m^3$ , averaged over four years.

![](_page_36_Figure_2.jpeg)

![](_page_37_Figure_0.jpeg)

**Fig. 10.** Monthly mean AOD and contributions to it from different aerosol species for China (dashed line) and India (solid line), averaged over four years.

![](_page_37_Figure_2.jpeg)

![](_page_38_Figure_0.jpeg)

![](_page_38_Figure_1.jpeg)

![](_page_38_Figure_2.jpeg)

![](_page_39_Figure_0.jpeg)

![](_page_39_Figure_1.jpeg)

**Fig. 12.** Monthly mean RH, 10-m wind speed and precipitation for China (dashed line) and India (solid line), averaged over four years.

4056

![](_page_40_Figure_0.jpeg)

![](_page_40_Figure_1.jpeg)

**Fig. 13.** Surface  $PM_{2.5}$  (in units  $\mu g/m^3$ ) and AOD (unitless) for India and China in the cases: no anthropogenic emissions (dotted line), 2006 emissions (dot-dot-dashed line), REAS reference scenario for 2020 (dashed line) and the -2% per year scenario (solid line), averaged over four years.