A HTAP multi-model assessment of the influence of regional anthropogenic emission reductions on aerosol direct radiative forcing and the role of intercontinental transport

Hongbin Yu\textsuperscript{1,2}, Mian Chin\textsuperscript{2}, J. Jason West\textsuperscript{3}, Cynthia S. Atherton\textsuperscript{4}, Nicolas Bellouin\textsuperscript{5}, Dan Bergmann\textsuperscript{4}, Isabelle Bey\textsuperscript{6}, Huisheng Bian\textsuperscript{7,2}, Thomas Diehl\textsuperscript{8,2}, Gerd Forberth\textsuperscript{5}, Peter Hess\textsuperscript{9}, Michael Schulz\textsuperscript{10}, Drew Shindell\textsuperscript{11}, Toshihiko Takemura\textsuperscript{12}, Qian Tan\textsuperscript{8,2}

1. Earth System Science Interdisciplinary Center, University of Maryland, College Park, Maryland, 20740, USA
2. Earth Science Directorate, NASA Goddard Space Flight Center, Greenbelt, Maryland, 20771, USA
3. Department of Environmental Sciences and Engineering, University of North Carolina, Chapel Hill, NC 27599, USA
4. Atmospheric Earth and Energy Division, Lawrence Livermore National Laboratory, Livermore CA 94551, USA
5. Met Office Hadley Centre, FitzRoy Road, Exeter Devon, EX1 3PB, UK
6. Center for Climate Systems Modeling, ETH Zurich, Universitätstrasse 16, 8092 Zürich, Switzerland
7. Joint Center for Earth Systems Technology, University of Maryland at Baltimore County, Baltimore, Maryland, 21228, USA
8. Universities Space Research Association, Columbia, Maryland, 21044, USA
9. Biological and Environmental Engineering, Cornell University, Ithaca, New York, 14850, USA
10. Meteorologisk Instituttt, Postboks 43 Blindern 0313, Oslo, Norway
11. NASA Goddard Institute for Space Studies, New York, New York, 10025, USA
12. Research Institute for Applied Mechanics, Kyushu University, Japan

Correspondence

Hongbin Yu
NASA GSFC Code 613
Greenbelt, MD 20771, USA
Hongbin.Yu@nasa.gov
301-614-6209
Abstract In this study, we assess changes of aerosol optical depth (AOD) and direct radiative forcing (DRF) in response to the reduction of anthropogenic emissions in four major pollution regions in the northern hemisphere by using results from 10 global chemical transport models in the framework of the Hemispheric Transport of Air Pollution (HTAP). The multi-model results show that on average, a 20% reduction of anthropogenic emissions in North America, Europe, East Asia and South Asia lowers the global mean AOD and DRF by about 9%, 4%, and 10% for sulfate, organic matter, and black carbon aerosol, respectively. The impacts of the regional emission reductions on AOD and DRF extend well beyond the source regions because of intercontinental transport. On an annual basis, intercontinental transport accounts for 10-30% of the overall AOD and DRF in a receptor region, with domestic emissions accounting for the remainder, depending on regions and species. While South Asia is most influenced by import of sulfate aerosol from Europe, North America is most influenced by import of black carbon from East Asia. Results show a large spread among models, highlighting the need to improve aerosol processes in models and evaluate and constrain models with observations.
1. Introduction

Anthropogenic aerosols make significant contributions to the global mean radiative forcing (RF) of climate [Forster et al., 2007] by scattering and absorbing solar radiation (so-called “aerosol direct effects”) [McCormick and Ludwig, 1967] and modifying cloud properties, amount, and evolution (collectively referred to as “aerosol indirect effects”) [Twomey, 1977; Gunn and Phillips, 1957; Albrecht, 1989]. RF is a measure of the change of net radiation (incoming minus outgoing) at the top of atmosphere (TOA), at the surface, or within the atmosphere, due to perturbations in atmospheric compositions or surface properties. On a global average basis, the sum of direct and indirect RF at TOA by anthropogenic aerosol is estimated to be -1.2 W m\(^{-2}\) [-2.4 to -0.6 W m\(^{-2}\)] (cooling) over the period of 1750-2000, which is significant compared to the positive (warming) forcing of +2.63 [±0.26] W m\(^{-2}\) by anthropogenic long-lived greenhouse gases over the same period [Forster et al., 2007]. In heavily polluted regions, aerosol cooling overwhelms greenhouse warming [Ramanathan et al., 2001; Li et al., 2010]. At the surface, aerosol RF can be much stronger than that at TOA because of aerosol absorption [Satheesh and Ramanathan, 2000]. Currently, uncertainties associated with aerosol RF make the largest contribution to the overall uncertainty in anthropogenic radiative forcing of climate [Forster et al., 2007]. Because of the significant role of aerosols in modulating the Earth’s radiative budget, it is necessary from both scientific and policy perspective to assess how emission changes associated with economic development and regional/national regulations will influence the aerosol radiative forcing.
The response of global aerosol RF to a fractional change of anthropogenic emissions would depend on the source locations, magnitude, and composition of emitted aerosols and aerosol precursors. While scattering aerosols like sulfate cause a cooling effect, strongly absorbing black carbon aerosols cause warming. Aerosol RF is also determined by several environmental factors such as surface albedo and meteorological conditions (in particular the amount and distribution of clouds and winds). With these differences, the aerosol RF per unit change of regional emissions would differ from one region to another. It is also important to note that the impact of a regional emission reduction is not confined to the region itself. Instead regional emission reductions can have far reaching impacts on RF in downwind regions, because of intercontinental transport of anthropogenic aerosols. Long-range transport has been observed by long-term surface monitoring networks [Prospero et al., 2003; VanCuren, 2003; Fischer et al., 2010], in-situ measurements from intensive field campaigns [Ramanathan et al., 2007; Clarke and Kapustin, 2010], and satellite observations [Yu et al., 2008; Rudich et al., 2008; Dirksen et al., 2009] backed by model simulations [Heald et al., 2006; Chin et al., 2007; Hadley et al., 2007]. Thus how a region is influenced by extraregional emissions is of particular importance for formulating an effective strategy for mitigating regional climate change and combating air pollution. The United Nations Economic Commission for Europe has established a Task Force on Hemispheric Transport of Air Pollutants (HTAP) to understand the growing body of scientific evidence of intercontinental transport and assess its impacts on air quality, climate, and ecosystems [HTAP, 2010].
Modeling studies can offer valuable insights into the relative significance of aerosols from different regions in influencing climate and important implications for formulating effective emission control strategies. Several recent studies have assessed how aerosols emitted in a region or from specific sectors could affect climate in downwind regions [e.g., Reddy and Boucher, 2007; Koch et al., 2007; Shindell et al., 2008a, 2008b]. Other studies have shown that large intermodel differences exist in aerosol lifetime cycle and radiative effect [Kinne et al., 2006; Schulz et al., 2006; Textor et al., 2006], which might undermine the robustness of the results from single model or very limited number of models. Complementary to the prior studies, we use in this study an ensemble of 10 global chemical transport or general circulation models to assess the change of global and regional aerosol optical depth (AOD) and direct radiative forcing (DRF) in response to 20% reductions of emissions from four major polluted regions in the Northern Hemisphere. These models conducted aerosol source-receptor (S/R) relationship experiments in the HTAP coordinated studies (HTAP 2010). These multi-model S/R experiments allow us to examine a probable range of contributions of intercontinental transport (ICT) relative to intra-regional emissions in determining regional AOD and DRF and help characterize the robustness of the results. Fry et al. [2012] conduct similar analysis of ozone radiative forcing due to 20% reductions of ozone precursor emissions using results from multiple HTAP models.

The rest of this paper is organized as follows. Section 2 describes the S/R model simulations and analysis methodology, including AOD from the HTAP anthropogenic S/R experiment used in this study and an estimate of the aerosol DRF. Section 3 presents
a multi-model analysis of the impacts of 20% reductions of regional anthropogenic emissions on global and regional AOD and DRF and the role of ICT relative to intra-regional emissions. This assessment does not address aerosol indirect effects, or warming effects resulting from BC deposition on snow and ice. We also neglect interactions of aerosols with thermal infrared radiation, as anthropogenic aerosols have relatively small size and their interactions with infrared radiation are minor. Finally, we neglect the radiative forcing of gas-phase components that may have been influenced by these emissions, which were modeled by Fry et al. (2012). Major conclusions from the analysis are summarized and discussed in Section 4.

2. Description of model simulations and analysis methodology

2.1. HTAP modeling experiments on S/R relationships

We use the results from an ensemble of 10 global chemical transport or general circulation models, listed in Table 1, to evaluate changes in AOD and DRF in response to a 20% reduction of anthropogenic emissions in four major pollution regions in the Northern Hemisphere under the framework of the HTAP S/R modeling experiments [Fiore et al., 2009; HTAP, 2010]. The S/R experiments include a baseline simulation and four perturbation simulations, for which each model submitted monthly AOD fields. For each model, the baseline simulation (SR1) is conducted using emissions and meteorology for 2001. Note that individual models use their own preferred anthropogenic and natural emissions. Each of the four perturbation runs (SR6) represents a 20% reduction in anthropogenic emissions (including biomass burning and n=both gas-phase and aerosol components) in one of four major pollution regions in Northern Hemisphere, namely
North America (NA), Europe (EU), East Asia (EA), and South Asia (SA), as illustrated in Figure 1. These perturbation model experiments are denoted as SR6NA, SR6EU, SR6EA, and SR6SA, respectively. Also shown in Figure 1 are regional and annual anthropogenic emissions of sulfur dioxide (SO$_2$), primary particulate organic matter (POM), and black carbon (BC) from 8 models (emissions for ECHAM5 and HadGEM2 were not archived and can not be retrieved for this analysis). Clearly anthropogenic emissions show large regional differences. For example, South Asia has the least SO$_2$ emissions that are a factor of 2-4 smaller than the other regions. East Asia has the largest BC emissions that are nearly double the emissions in Europe or South Asia and more than triple the emissions in North America. The regional differences in emissions in combination with the proximity of regions (e.g., the SA region is adjacent to and downwind of the EU region), ICT pathways, and removal/production mechanisms during the transport, would determine the relative roles of ICT and regional emissions, as we will discuss later in the paper.

2.2 AOD from HTAP S/R experiments

The major HTAP model outputs used in this study are monthly average AOD (τ) at 550 nm for sulfate (SO$_4$), POM, and BC from individual models. Figure 2 shows annual average AOD in ambient conditions for a combination of SO$_4$, POM, and BC from 9 models. ECHAM5 simulations are not shown here because the model calculates AOD for dry sulfate, POM, and BC. Although the model also provides water optical depth associated with total aerosol (e.g., due to aerosol humidification), it is impossible, without further uncertain assumptions, to partition the water optical depth into that
associated with individual aerosol components. Note that these AOD outputs include both anthropogenic and natural contributions (e.g., DMS derived sulfate AOD and biomass burning AOD, among others). The AOD distributions clearly show several hotspots representing well-known industrial pollution regions (e.g., East and South Asia, Western Europe, and eastern US) and biomass burning regions (e.g., equatorial Africa in DJF, South America and southern Africa in JJA and SON). In Table 2, the global annual mean AOD is 0.0335 ± 0.0134 (average ± standard deviation of the 9 models), 0.0108 ± 0.0047, and 0.0022 ± 0.0024 for SO$_4$, POM, and BC, respectively. Clearly AOD shows large model diversity with the standard deviation equivalent to about half of the multi-model average. These global annual means are lower than those from multiple model simulations with harmonized emissions under the framework of the Aerosol Comparisons between Observations and Models (AeroCom) [Schulz et al., 2006] by 10%, 28%, and 8% for sulfate, POM, and BC, respectively.

Figure 3(a) shows the 9-model median AOD for sulfate, POM, and BC combined as derived from the SR1 baseline simulations and its comparisons with anthropogenic AOD derived from MODIS over-ocean measurements using two different methods described by Yu et al. [2009] (Figure 3b, denoted as MODIS-YU09) and Bellouin et al. [2008] (Figure 3c, denoted as MODIS-BE08). MODIS-YU09 anthropogenic AOD is derived from MODIS over-ocean retrievals of AOD and fine-mode fraction (FMF) using the representative FMF values for individual aerosol types, which are determined from MODIS observations in selected regions where the specific aerosol type predominates [Yu et al., 2009]. MODIS-BE08 is derived from the same MODIS measurements, but
with the use of prescribed, in situ measurement-based thresholds of FMF for different aerosol types (which are different from those derived from MODIS observations as in Yu et al., 2009) aided by the satellite observed absorbing aerosol index to separate anthropogenic aerosol from mineral dust and sea salt [Bellouin et al., 2008]. Deriving anthropogenic AOD from satellite measurements over land is not currently feasible. As shown in Figure 3, anthropogenic AOD of MODIS-YU09 is generally larger than MODIS-BE08 AOD in northern hemispheric mid-latitudes. In the tropics, MODIS-BE08 AOD is somewhat higher than MODIS-YU09. Except for trans-Pacific transport in spring, the cross-ocean transport from major industrial pollution and biomass burning regions is generally more extensive in the HTAP model simulations than both MODIS-based estimates. In particular, the models simulated significant cross-Atlantic transport of tropical African smoke to South America in boreal winter, which is not clearly seen in the MODIS-based estimates. Figure 4 further compares zonal variations of HTAP seasonal AOD for the sulfate, POM and BC mixture against that of the MODIS anthropogenic AOD over the ocean in mid-latitudes (20-60°N) of the Northern Hemisphere, where the major intercontinental transport paths occur. Significant regional and seasonal differences exist between the HTAP models and MODIS observations and between MODIS-YU09 and MODIS-BE08. In the North Atlantic, the MODIS anthropogenic AOD from both methods is often near the low bound of 9 HTAP models. In the North Pacific, MODIS-YU09 anthropogenic AOD generally agrees well with the median of HTAP models. On the other hand, MODIS-BE08 is consistently lower than MODIS-YU09 and generally entails the low bound of the HTAP model simulations, particularly in the northeastern Pacific. The differences between the two MODIS-based estimates of anthropogenic AOD
result from the differences in specifics of the approaches, as discussed in Yu et al. [2009]. The pronounced differences between the two MODIS-based estimates of anthropogenic AOD manifest the urgent need to develop independent approaches based on other satellite measurements, and more importantly, to conduct in situ measurements of anthropogenic AOD over large scales [Yu et al., 2009].

2.3 Estimate of the aerosol direct radiative forcing

Because the aerosol direct radiative forcing or the aerosol optical properties (i.e., single-scattering albedo and asymmetry factor) for calculating the forcing are not archived under the HTAP S/R experiment protocol, we estimate here the aerosol DRF for each model and component (i.e., sulfate, POM, or BC) as follows:

\[
\text{DRF}_i(x, y, t) = AOD_i(x, y, t) \cdot \text{NDRF}_i(x, y, t) \quad (i = \text{SO}_4, \text{POM}, \text{and BC})
\]

where \(x, y, \text{ and } t\) represents longitude, latitude, and month, respectively, and NDRF is the normalized DRF with respect to AOD at 550 nm [Zhou et al., 2005]. In this study we derive monthly average NDRF\(_i\) over each model grid cell by dividing the DRF by the AOD, using 2001 monthly average AOD and direct radiative forcing calculated from the Goddard Chemistry Aerosol Radiation and Transport (GOCART) model [Chin et al., 2002]. We then apply this NDRF\(_i\) to the monthly average component AOD\(_i\) from the other models in the HTAP S/R experiments. We also estimate DRF for an external mixture of SO\(_4\), POM and BC by simply summing up the DRF for individual components. Note that RF for an internal mixture could differ significantly from that for an external mixture [Jacobson, 2001; Chung and Seinfeld, 2002]. Although DRF does not increase with AOD in a fully linear manner over the whole range of AOD, the use of
the above linear formula to derive DRF and its change in response to a 20% reduction of emissions would not introduce large uncertainties with respect to regional differences in DRF [Zhou et al., 2005; Anderson et al., 2005].

The GOCART model currently prescribes particle size distributions and refractive indices for individual components based on the Optical Properties of Aerosols and Clouds (OPAC) database [Hess et al., 1998]. Aerosol properties such as AOD, single scattering albedo, and asymmetry factor are then calculated using the Mie code [Chin et al., 2002, 2009]. These aerosol optical properties along with surface albedos and cloud fields from the Goddard Earth Observing System (GEOS) Data Assimilation System - Version 4 are then used to drive the NASA Goddard radiative transfer model [Chou et al., 1998]. The time step for the radiative transfer calculations is 30 min, which can adequately capture the dependence of DRF on solar zenith angle [Yu et al., 2004]. For a specific component (e.g., SO$_4$, POM, or BC), DRF is calculated as the difference of net downward radiative flux between a radiative transfer calculation including all aerosol components and one with the specific component excluded. This DRF is therefore different from the DRF estimated in Forster et al. [2007], where the reference is pre-industrial aerosols. In this study, the GOCART DRF is calculated for solar radiation only and averaged over a 24-hour period. GOCART calculations of DRF have been evaluated against remote sensing measurements and other model simulations [e.g., Yu et al., 2004; Yu et al., 2006].

Table 3 lists the annual mean NDRF in the four defined regions and globally. For purely scattering SO$_4$, NDRF is nearly the same at the top of atmosphere (TOA) and at the
surface. Because POM is partially absorbing in the UV range [Chin et al., 2009], the surface NDRF is more negative than the TOA NDRF. BC aerosol is strongly absorbing over the whole solar spectrum, thus its TOA forcing is positive (i.e., TOA warming), while the surface forcing is strongly negative (i.e., surface cooling). Clearly BC aerosol is much more effective in interacting with solar radiation than SO₄ and POM, although it is typically associated with lower AODs. For a specific component, the difference in NDRF among regions is generally within 20-30%, due to the combined effects of differences in solar zenith angle, surface albedo, and cloud fields [Yu et al., 2006]. The global annual mean all-sky TOA NDRF is -24.2, -30.0, and +85.9 Wm⁻²τ⁻¹ for SO₄, POM, and BC, respectively. These values fall within the ranges reported in the literature, i.e., -10 ~ -32 Wm⁻²τ⁻¹ for SO₄, -5 ~ -38 Wm⁻²τ⁻¹ for POM, and +22 ~ +216 Wm⁻²τ⁻¹ for BC [CCSP, 2009].

To apply the GOCART-based monthly NDRF to the HTAP models, we regrid the component AOD from each model to a unified spatial resolution of 2°x2.5° and use eq. (1) to estimate the corresponding DRF. For the SR1 experiment and on a global and annual basis, the 9 models give the mean all-sky TOA DRF of -0.67 ± 0.28, -0.26 ± 0.13, and +0.22 ± 0.13 Wm⁻² for SO₄, POM, and BC, respectively. Correspondingly, the respective DRF at the surface and in all-sky (including both cloudy and cloud-free) condition is -0.64 ± 0.26, -0.37 ± 0.19, and -0.43 ± 0.22 Wm⁻². When the anthropogenic fraction of AOD, e.g., 55%, 53%, and 100% respectively for SO₄, POM, and BC on a global and annual average basis [Schulz et al., 2006] is considered, our DRF estimates shown above are in good agreement with anthropogenic TOA DRF of -0.35 ± 0.15 Wm⁻².
The aerosol direct radiative forcing exhibits large regional and seasonal variations, as shown in Figure 5 for the all-sky DRF and for the external mixture of SO$_4$, POM, and BC, as derived from the 9-model SRI simulations. Over major industrial pollution regions in northern hemispheric mid-latitudes, aerosol DRF can be as large as -6 Wm$^{-2}$ at TOA and -25 Wm$^{-2}$ at the surface. The forcing is greater in JJA and MAM than in DJF and SON. Similar magnitudes of DRF are seen in the tropics where biomass burning smoke is dominant. The seasonal variations of DRF reflect the peak seasons of biomass burning, e.g., DJF in the Sahel and JJA and SON in southern Africa and South America.

We note that the use of GOCART-based NDRF is likely to understate the model diversity in DRF, in comparison to that derived from full radiative transfer calculations by participating models. Although other aerosol models currently use the OPAC to characterize aerosol microphysical and optical properties just like GOCART, models can differ substantially in aerosol vertical distributions, meteorological fields (such as relative humidity, cloud distributions, and surface albedos), and radiative transfer schemes. These differences combined may lead to large model differences in DRF. This could particularly be the case for all-sky TOA DRF by BC, which depends strongly on the vertical distributions of aerosol and clouds, as evidenced by the reported wide range of DRF [CCSP, 2009].
3. Influences of regional anthropogenic emission reductions on AOD and DRF

In this section we present how the 20% reduction of regional anthropogenic emissions changes AOD and DRF by analyzing differences between a set of SR6 experiments (SR6NA, SR6EU, SR6EA, and SR6SA) and the SR1 experiment. We examine first the changes in global mean AOD and DRF and then the spatial extents of AOD and DRF changes resulting from regional anthropogenic emission reductions. The relative roles of ICT and regional emissions are assessed with a concept of the relative annual intercontinental response (RAIR) (as defined in HTAP, 2010). For the reasons discussed earlier in 2.2, we exclude ECHAM5 results in the analysis of AOD and DRF magnitude. Because RAIR is less sensitive to water optical depth than AOD is, we include ECHAM5 results in the RAIR analysis.

3.1. Global mean

Just as aerosol radiative forcing varies regionally, the effects of changes in emissions of aerosols and their precursors on global mean DRF also varies with region. Tables 4 and 5 list respectively changes of global annual average AOD and all-sky DRF in response to the 20% reduction of anthropogenic emissions in the four HTAP anthropogenic source regions. The combined impact of the 20% reduction of emissions in all of the four regions is to decrease AOD and DRF by about 9%, 4%, and 10% for sulfate, POM, and BC, respectively. Relative contributions from individual regions vary considerably. For sulfate, the change of global average AOD and DRF due to the reduction of SO$_2$ emissions in South Asia is substantially smaller than that due to the emission reductions.
from the other regions. This is mainly because SO$_2$ emissions in the SA region are much smaller than those in other regions (see Figure 1). For POM, the reductions of global mean AOD and DRF due to the regional emission reduction are generally consistent with regional differences in primary POM emissions. For BC, the reduction of emissions in East Asia makes the largest contribution to the change of global average DRF, mainly because of the highest BC emissions among the four regions. For the external mixture of SO$_4$, POM, and BC, the 20% reductions in the anthropogenic emissions in the four regions collectively yield the respective reductions of 0.0035, 45.5 mWm$^{-2}$, (1 mWm$^{-2}$ = 10$^{-3}$ Wm$^{-2}$) and 104.1 mWm$^{-2}$ for global mean AOD, all-sky DRF at TOA and at the surface, which represents respectively about 8%, 6% and 7% reduction from the baseline simulation. Note that uncertainties are larger than given by the standard deviation in Table 5, due to the use of NDRF from a single model, as discussed earlier.

Will the same amount of emission reduction in different regions introduce the same change in global mean aerosol radiative forcing? To address this question, we calculate the forcing efficiency with respect to emissions by normalizing the global annual mean DRF difference between the SR6 and SR1 runs by the change (20%) in regional emissions for each model, as shown in Table 6. For SO$_4$ and POM, we use the emissions for SO$_2$ and primary POM respectively to calculate the forcing efficiency. These derived forcing efficiency numbers should be considered as rough estimates, because some models with fully coupled chemistry include changes of SO$_4$ resulting from the reductions of other anthropogenic emissions (Fry et al., 2012) and a fraction of POM is secondary aerosol formed from volatile organic carbon emissions. Clearly the forcing
efficiency for BC is an order of magnitude higher than that for SO$_4$ and POM, suggesting that BC emission control is a candidate for producing a large, near-term cooling that counters the warming of greenhouse gases. Despite wide model differences in the absolute value of forcing efficiency, several regional dependences appear to be rather robust among the models. As shown in Figure 6, all eight models consistently yield a smaller sulfate forcing efficiency for EA emissions than for EU emissions, which is consistent with the oxidant limitation in EA that reduces the efficiency of SO$_2$-to-sulfate transformation [Koch et al., 2007]. The models also consistently give the largest BC forcing efficiency for EU emissions. These results may have important implications for regional emission controls and their influences on global climate, which will be discussed in section 4.

3.2. Spatial extents and the role of ICT

The spatial extents of the AOD and DRF due to the reduction of regional emissions and the role of aerosol ICT are revealed by differencing SR6 and SR1 experiments. Figure 7 shows the 9-model annual average changes of AOD and all-sky DRF by the external mixture of SO$_4$, POM, and BC, resulting from the 20% reduction of anthropogenic emissions over the four regions. Clearly emissions from North America, Europe, and East Asia exert significant DRF on intercontinental and even hemispheric scales. Emissions from South Asia have relatively small impacts mainly over the Indian Ocean and the tropical Pacific. The influences of regional emission reduction on AOD and DRF also depend on season. As an example, Figure 8 shows seasonal variations of changes of the all-sky surface DRF between SR1 and SR6. Similar seasonal variations occur for AOD,
all-sky DRF at TOA and clear-sky DRF, which will not be discussed separately. For
North America, Europe, and East Asia, the 20% reduction of emissions in each region
decreases the direct forcing (i.e., less negative) by a larger amount and over more
extensive areas in summer and spring than in winter and fall. For comparison, the
influence of the 20% reduction of emissions in South Asia shows relatively small
seasonal variations.

To quantify the role of aerosol ICT in affecting regional climate forcing, we adopt a
concept of the RAIR as defined in HTAP [2010]. For AOD, RAIR in a receptor region i
is expressed as follows:

\[
RAIR_i = \frac{\sum_j \delta AOD_{ji}}{\delta AOD_{ii} \sum_j \delta AOD_{ji}} \quad (j \neq i)
\]

where index j represents a source region. \(\delta AOD_{ii}\) represents a change of AOD in the
receptor/domestic region i due to the emission reduction in the region itself and \(\delta AOD_{ji}\)
represents a change of AOD in the receptor i induced by the emission reduction in a
source region j outside of the receptor (or foreign region). Similarly, RAIR can be
defined for DRF, near-surface concentration, and surface deposition. By definition, RAIR
in a receptor region represents the percentage contribution of the intercontinental
transport of foreign emissions relative to the sum of foreign and domestic emissions.

We calculate the RAIR for AOD and DRF for individual models and then obtain multi-
model averages, as shown respectively in Figure 9 and 10 with the columns representing
10-model (including ECHAM5) average and error bars indicating ±1 standard deviation of the model simulations. Clearly values of RAIR depend on both region and component. For all regions and components, the import from intercontinental transport is significant but local emissions remain the main contributor, with RAIR ranging from 10 to 30%. South Asia is most influenced by the import of sulfate aerosol, and North America is most influenced by the import of BC, followed by POM. For sulfate, POM, and BC combined, South Asia DRF is most strongly influenced by foreign sources, and North America and Europe is the least. These rankings also reflect the strength of local emissions relative to world emissions, as discussed in section 2. Interestingly, RAIR values for AOD and DRF of sulfate (10-30%) are consistently smaller than that for aerosol column loading (i.e., 24-37%) [see Table 4.4 in HTAP 2010]. For POM, RAIR values (16-18%) for AOD and DRF in NA and EU are also somewhat smaller than the corresponding RAIR values (21-23%) for the column mass loading. These differences probably stem from transported sulfate and POM aerosols experiencing lower relative humidity at high altitude (resulting in lower optical depth) than local aerosols, which generally remain at lower altitudes. Finally, model variability is very large, highlighting the significant uncertainties in modeling aerosol processes.

The contributions of emission reductions in a receptor region depend on source regions. Table 7 lists the response of annual mean AOD and all-sky DRF in receptor regions to a 20% reduction of anthropogenic emissions in the four source regions. Here we combine SO$_4$ and POM together, in which SO$_4$ makes a dominant contribution of 69-94%, depending on regions. BC is listed separately from the SO$_4$ and POM combined, because
strongly absorbing BC has different effects on solar radiation than do weakly absorbing
sulfate and organic matter. Not surprisingly, the influences on AOD and DRF from the
reduction of domestic emissions is about an order of magnitude larger than from any
foreign source region. Foreign source regions also differ in contributing to the total
import for a specific receptor region. The intercontinental transport of EA emissions
accounts for the largest fraction of total import in NA, i.e., 60% and 70% for SO$_4^{+}$POM,
and BC, respectively. The NA emissions make the largest contribution (57%) to the total
import of SO$_4^{+}$POM into EU. For BC aerosols imported to EU, on the other hand, the EA
collection of 45% exceeds the NA contribution of 35%, presumably because of much
higher BC emissions in EA. The imported BC to EA comes mostly (65%) from SA,
followed by those from EU (e.g., 30%). For the import of SO$_4^{+}$POM to EA, SA and EU
emissions make quite comparable contributions, which is more than a factor of 3 larger
than NA emissions. In SA, the imported SO$_4^{+}$POM is significant, with RAIR of 24%, of
which 66% comes from EU, followed by 28% from EA. For BC aerosols in SA, the
import from EA and EU emissions contributes more or less equally (e.g., 48% and 44%).

Our multi-model estimates of the relative contributions of local emissions and ICT import
are generally consistent with results in literature. In NA, we estimate that local emissions
account for about 81% of anthropogenic AOD, which is higher than 66% estimated by a
modeling study [Leibensperger et al., 2012]. This difference would be narrowed if
emissions in regions other than the four defined regions of this study are taken into
account. Yu et al. [2012a] estimate based on satellite measurements that the imported
pollution via trans-Pacific transport results in a clear-sky TOA direct forcing of -0.49
Wm$^2$ over NA, which accounts for about 11.5% of total anthropogenic aerosol direct forcing. For comparison, our estimated contribution by ICT import to the direct forcing over NA is 14.7% in this study. The BC fractional contributions from different source regions as estimated in this study also agree well with a model estimate [Reddy and Boucher, 2009, referred to as RB09]. RB09 estimated that over North America, emissions from East and South Asia contribute to 18% and 6% of the total BC burden, respectively, which are somewhat smaller than our corresponding estimate of 23% and 8%. The contribution of North America emissions to BC burden in Europe was estimated at 5% in RB09, which is also somewhat smaller than our estimated 8%. RB09 estimated that the local BC emissions over South and East Asia accounts for more than 80% of the BC burden, which agrees well with our estimate of 82% and 84%. These comparisons show how the estimated significance of ICT import in this study agrees broadly with results in the literature, as these studies have defined source regions somewhat differently.

4. Conclusions and discussion

We have assessed impacts of a 20% reduction of anthropogenic emissions in North America, Europe, East Asia, and South Asia on the aerosol optical depth and direct radiative forcing by using results from 10 global chemical transport models. On the basis of the multi-model average, a 20% reduction of anthropogenic emissions in the four regions combined lowers the global mean AOD and DRF by about 9%, 4%, and 10% for sulfate, organic matter, and black carbon aerosol, respectively. Despite the considerable model-to-model differences in the magnitude of forcing efficiency (direct radiative forcing per unit emissions), the models appear to consistently give a lower sulfate forcing
efficiency for SO$_2$ emissions from EA than that from EU, and the largest BC forcing efficiency for the EU emissions. These results may suggest that reducing BC emissions in EU would be relatively the most efficient way to mitigate the warming effect of BC aerosol, not considering cost and feasibility. However, a larger potential for BC reductions is available in East Asia. Reducing SO$_2$ emissions decreases the cooling effect of aerosols that counteracts the greenhouse warming; but the impact of reducing EA SO$_2$ is less pronounced (ca. 25%) than reducing EU, NA, or SA SO$_2$ by the same amount. However, such implications need to be further explored by taking into account a variety of aerosol impacts on climate through modifying cloud microphysics, changing atmospheric circulations, and altering the snow albedo. The air quality implication of any emission control should also be considered.

Our multi-model simulations of source-receptor relationships show that the impacts of the regional emission reductions are not confined to the region itself because of aerosol intercontinental transport. On an annual basis, intercontinental transport accounts for 10-30% of the overall AOD or DRF in a receptor region, compared to the influence of both regional emissions and intercontinental transport, depending on regions and species. While South Asia is most influenced by import of sulfate aerosol from Europe, North America is most influenced by import of black carbon aerosol from East Asia. Given that BC deposition may accelerate the melting of snow in the Sierra Nevada and cause water supply shortage in summer in the western U.S. [Hadley et al., 2010], the region may likely benefit from a future control of BC emissions in Asia.
The multi-model assessment shows large differences between models in the impacts of emission reductions and the role of intercontinental transport, which highlights a need for improving models and developing observational databases for evaluating and constraining models. Quantifying anthropogenic AOD distributions from satellite measurements remains challenging, in particular over land. Current estimates of anthropogenic AOD based on total AOD and fine-model fraction measurements from MODIS are only feasible over ocean and subject to large uncertainties [Yu et al., 2009; Bellouin et al., 2005, 2008]. Such an estimate would be better constrained by a synergistic use of aerosol microphysical measurements as provided by other satellite sensors [Yu et al., 2012b and references therein]. Large-scale in situ measurements of anthropogenic AOD will be essential for evaluating satellite observations and model simulations. From the perspective of model improvements, efforts should focus on not only emission inventories [Textor et al., 2007] but also a variety of atmospheric processes that determine the atmospheric evolution of aerosols, such as parameterization of dry deposition and wet scavenging [Prospero et al., 2011]. Given that the aerosol intercontinental transport and its influences involve a wide span of scales, it is necessary to develop modeling systems that link the local, regional, intercontinental, and global scales.

**Acknowledgements:** HY was supported by NASA grant NNX11AH66G, managed by Richard Eckman. MC and HB were supported by NASA Modeling, Analysis, and Prediction program managed by David Considine. NB thanks Shekar Reddy, formerly with the Met Office Hadley Centre, for carrying out the HadGEM2 simulations. We are
grateful to many colleagues, including Frank Dentener and Bill Collins for helpful
comments. The HTAP experiments used in this study were organized by Martin Schultz,
Arlene Fiore, Kees Cuvelier, Frank Dentener, Christiane Textor, Terry Keating, and
Andre Zuber.

References

245, 1227-1230.

Anderson, T.L., et al. (2005), An ”A-Train” strategy for quantifying direct aerosol

Bellouin, N., O. Boucher, J. Haywood, and M. S. Reddy (2005), Global estimate of
aerosol direct radiative forcing from satellite measurements, *Nature*, 438, 1138-
1141, doi:10.1038/nature04348.

Bellouin, N., J. Rae, A. Jones, C. Johnson, J. Haywood, and O. Boucher (2011), Aerosol
forcing in the Climate Model Intercomparison Project (CMIP5) simulations by
HadGEM2-ES and the role of ammonium nitrate, *J. Geophys. Res.*, 116, D20206,

Bellouin, N., A. Jones, J. Haywood, and S. A. Christopher (2008), Updated estimate of
aerosol direct radiative forcing from satellite observations and comparison against
the Hadley Centre climate model, *J. Geophys. Res.*, 113, D10205,

Sensitivity of aerosol optical thickness and aerosol direct radiative effect to relative

CCSP (2009), Atmospheric Aerosol Properties and Climate Impacts, A Report by the
*U.S. Climate Change Science Program and the Subcommittee on Global Change*


Schulz, M. (2007), Constraining model estimates of the aerosol radiative forcing, Thèse d'Habilitation à Diriger des Recherches, Université Pierre et Marie Curie, Paris VI.


Table Captions

Table 1. List of 10 models that participated in HTAP aerosol S/R experiments and are used in this analysis.

Table 2. Comparisons of AOD at 550 nm ($\tau$) from 9 HTAP models in this study with that inferred from Schulz et al. (2006). For SO$_4$ and POM, AOD is inferred by using anthropogenic AOD and anthropogenic fraction of present day AOD. For BC, we infer AOD from the anthropogenic absorptive AOD reported in Schulz et al. [2006] by assuming that all BC is anthropogenic and BC has a single scattering albedo of 0.2 at 550 nm.

Table 3. Annual mean aerosol direct radiative forcing normalized by AOD at 550 nm (NDRF, Wm$^{-2}\tau^{-1}$) at top of atmosphere (TOA) and surface for SO$_4$, POM, and BC in the four source regions (NA, EU, EA, SA) and globe, which is derived from GOCART simulated monthly AOD and DRF for 2001.

Table 4. Change of global annual mean AOD ($\tau \times 1000$, mean ± std. dev) in response to the 20% reduction of anthropogenic emissions in 4 source regions (SR6 - SR1) of 9 HTAP models.

Table 5. Change of global annual mean aerosol DRF (unit: mWm$^{-2}$, mean ± standard deviation, 1 mWm$^{-2} = 0.001$ Wm$^{-2}$) in response to the 20% reduction of anthropogenic emissions in 4 source regions as derived from analysis of SR1 and SR6 runs of 9 HTAP models.

Table 6. Global annual average TOA all-sky forcing efficiency relative to emissions from the source regions (unit: mWm$^{-2}$ per Tg) as derived from 8 HTAP models. The forcing efficiency for SO$_4$ and POM is calculated with respect to SO$_2$ and primary POM emissions, although some models with fully coupled chemistry include changes of SO$_4$ resulting from the reductions of other emissions and a fraction of POM is secondary aerosol produced from a variety of volatile organic carbon.
Table 7: Response of annual AOD and all-sky DRF (mean ± std. dev.) in the receptor regions (columns) to 20% reductions of anthropogenic gas and aerosol emissions in the source regions (rows) estimated from 9 HTAP models. Percentage contributions of individual source regions are shown in parentheses for AOD (corresponding percentages for DRF are only slightly different and hence not shown). Contributions of a region on itself (termed “domestic”) are shown in bold. Here BC is listed separately from SO$_4$ and POM combined (in which SO$_4$ makes a dominant contribution of 69-94%, depending on region), because of their distinct effects on solar radiation. For SO$_4$+POM, surface DRF is slightly greater than TOA DRF because of weak absorption and not shown. A positive change of DRF for SO$_4$+POM suggests a weakened cooling effect by the emission reduction. A BC emission reduction introduces a negative change of TOA DRF (a weakened warming effect) and a positive change of surface DRF (a weakened cooling effect).
**Figure Captions**

**Figure 1:** Illustration of four HTAP defined regions for examining the source-receptor relationships for anthropogenic aerosols: North America (NA, 15°-55°N, 60°W-125°W), Europe (EU, 25°-65°N, 10°W-50°E), East Asia (EA, 15°-50°N, 95°-160°E), and South Asia (SA, 5°-35°N, 50°-95°E). Regional and annual anthropogenic emissions of SO$_2$, primary POM, and BC from 8 HTAP models (excluding HADGEM2 and ECHAM5) are shown in bar charts, with error bar indicating the range of 8 models.

**Figure 2:** Annual average AOD for the external mixture of sulfate, POM, and BC simulated by baseline runs of 9 HTAP models.

**Figure 3:** Comparison of HTAP SR1 9-model median AOD for the external mixture of sulfate, POM, and BC (a) with MODIS-derived anthropogenic AOD over ocean as described in Yu et al. [2009] (b) and Bellouin et al. [2008] (c).

**Figure 4:** Comparisons of zonal variations of seasonal and 20°N-60°N average AOD for sulfate, POM, and BC combined as simulated by HTAP models (black line for median and shaded area for the range of 9 HTAP models) with the MODIS-derived over-ocean anthropogenic AOD (red line for Yu et al., 2009 and blue line for Bellouin et al., 2008).

**Figure 5:** Seasonal variations of 9-model average TOA DRF (a), and surface DRF (b) in all-sky conditions (Wm$^{-2}$) for the external mixture of sulfate, POM, and BC as derived from the HTAP baseline simulations (SR1).

**Figure 6:** Global annual mean forcing efficiency (unit: mWm$^{-2}$ per Tg) for sulfate (a) and BC (b) with respect to regional (denoted as NA, EU, EA, and SA) emissions as simulated by 8 HTAP models.
Figure 7: Annual average AOD (x100, top panel) and DRF (mWm$^{-2}$) at the top of atmosphere (TOA) (middle panel) and at the surface (bottom panel) in all-sky conditions resulting from 20% reductions of regional anthropogenic emissions over North America (NA), Europe (EU), East Asia (EA), and South Asia (SA), respectively. A positive value for DRF represents the reduced aerosol direct radiative forcing resulting from the reduction of emissions. Each of the four source regions is overlaid on corresponding SR6xx-SR1 maps where xx represents NA, EU, EA, and SA. Individual aerosol components are assumed to be mixed externally. The results are based on 9 HTAP models.

Figure 8: Seasonal variations of changes in the all-sky DRF (mWm$^{-2}$) at surface (SR6-SR1) due to 20% reduction of regional anthropogenic emissions derived from 9 HTAP models. Each of the four source regions is overlaid on corresponding SR6xx-SR1 maps where xx represents NA, EU, EA, and SA.

Figure 9: Ten-model (including ECHAM5) derived average relative annual intercontinental response (RAIR) for aerosol optical depth (AOD) in four receptor regions, by chemical component. Standard deviations, as indicated by error bars, reflect the model variability in simulating aerosol transport.

Figure 10: same as Figure 9 but for TOA all-sky DRF.
<table>
<thead>
<tr>
<th>Model</th>
<th>Model version</th>
<th>Investigator(s)</th>
<th>Resolution (lon. x lat.)</th>
<th>Major Reference(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>CAM-CHEM</td>
<td>v3311m13</td>
<td>P. Hess</td>
<td>2.5°x1.875°</td>
<td>Pfister et al., 2008</td>
</tr>
<tr>
<td>CAM-CHEM</td>
<td>v3514</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ECHAM5-HAMMOZ</td>
<td>v21</td>
<td>I. Bey, G. Forberth</td>
<td>2.813°x2.813°</td>
<td>Pozzoli et al., 2008a and 2008b</td>
</tr>
<tr>
<td>GISS PUCCINI</td>
<td>modelEaer</td>
<td>D. Shindell</td>
<td>5°x4°</td>
<td>Koch et al., 2005;</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Shindell et al., 2006</td>
</tr>
<tr>
<td>GMI</td>
<td>v02a</td>
<td>H. Bian</td>
<td>2.5°x2°</td>
<td>Bian et al., 2009</td>
</tr>
<tr>
<td>GOCART</td>
<td>v4p2</td>
<td>M. Chin, T. Diehl</td>
<td>2.5°x2°</td>
<td>Chin et al., 2002,</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2007, 2009</td>
</tr>
<tr>
<td>HADGEM2</td>
<td>A-v01</td>
<td>N. Bellouin</td>
<td>1.875°x1.25°</td>
<td>Bellouin et al., 2011</td>
</tr>
<tr>
<td>INCA</td>
<td>v2</td>
<td>M. Schulz</td>
<td>3.75°x2.5°</td>
<td>Schulz, 2007;</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Hauglustaine et al.,</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2004; Textor et al.,</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2006</td>
</tr>
<tr>
<td>LLNL IMPACT</td>
<td>T5a</td>
<td>D. Bergmann, C. S. Atherton</td>
<td>2.5°x2°</td>
<td>Rotman et al., 2004.</td>
</tr>
<tr>
<td>SPRINTARS</td>
<td>v356</td>
<td>T. Takemura</td>
<td>1.125°x1.125°</td>
<td>Takemura et al., 2005</td>
</tr>
</tbody>
</table>

**Table 1.** List of 10 models that participated in HTAP aerosol S/R experiments and are used in this analysis

**Acronyms for model names:**
- CAM-CHEM: Community Atmospheric Model – Chemistry version (NCAR, USA)
- ECHAM5-HAMMOZ: Max-Planck Institute for Meteorology Hamburg Climate Model - version 5 with Hamburg Aerosol Model and MOZart chemistry sub-Module
- GISS PUCCINI: Goddard Institute for Space Studies, Physical Understanding of Composition-Climate Interactions and Impacts model (NASA GISS, USA)
- GMI: Global Modeling Initiative (NASA GSFC, USA)
- GOCART: Goddard Chemistry Aerosol Radiation and Transport (NASA GSFC, USA)
- HADGEM2: Hadley Centre Global Environment Model version 2 (Met Office, UK)
- INCA: Interaction of Chemistry and Aerosol (laboratoire des Sciences du Climat et de l'Environnement, France)
- LLNL IMPACT: Integrated Massively Parallel Atmospheric Chemical Transport model (Lawrence Livermore National Laboratory, USA)
- SPRINTARS: Spectral Radiation-Transport Model for Aerosol Species (Kyushu University, Japan)
Table 2. Comparisons of AOD at 550 nm (τ) from 9 HTAP models in this study with that inferred from Schulz et al. (2006). For SO$_4$ and POM, AOD is inferred by using anthropogenic AOD and anthropogenic fraction of present day AOD. For BC, we infer AOD from the anthropogenic absorptive AOD reported in Schulz et al. [2006] by assuming that all BC is anthropogenic and BC has a single scattering albedo of 0.2 at 550 nm.

<table>
<thead>
<tr>
<th></th>
<th>This study</th>
<th>Schulz et al. [2006]</th>
</tr>
</thead>
<tbody>
<tr>
<td>τ$_{SO4}$</td>
<td>0.0335 ± 0.0134</td>
<td>0.0345 ± 0.0164</td>
</tr>
<tr>
<td>τ$_{POM}$</td>
<td>0.0108 ± 0.0047</td>
<td>0.0151 ± 0.0094</td>
</tr>
<tr>
<td>τ$_{BC}$</td>
<td>0.0022 ± 0.0010</td>
<td>0.0024 ± 0.0010</td>
</tr>
</tbody>
</table>
Table 3. Annual mean aerosol direct radiative forcing normalized by AOD at 550 nm (NDRF, Wm\(^{-2}\)sr\(^{-1}\)) at top of atmosphere (TOA) and surface for SO\(_4\), POM, and BC) in the four source regions (NA, EU, EA, SA) and globe, which is derived from GOCART simulated monthly AOD and DRF for 2001.

<table>
<thead>
<tr>
<th>Sky Condition</th>
<th>Region</th>
<th>NDRF(_{SO4})</th>
<th>NDRF(_{POM})</th>
<th>NDRF(_{BC})</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>TOA</td>
<td>Surface</td>
<td>TOA</td>
</tr>
<tr>
<td>All Sky</td>
<td>NA</td>
<td>-24.9</td>
<td>-24.5</td>
<td>-28.4</td>
</tr>
<tr>
<td></td>
<td>EU</td>
<td>-21.1</td>
<td>-20.4</td>
<td>-22.6</td>
</tr>
<tr>
<td></td>
<td>EA</td>
<td>-21.4</td>
<td>-21.2</td>
<td>-25.4</td>
</tr>
<tr>
<td></td>
<td>SA</td>
<td>-24.9</td>
<td>-25.1</td>
<td>-28.7</td>
</tr>
<tr>
<td></td>
<td>globe</td>
<td>-24.2</td>
<td>-24.1</td>
<td>-30.0</td>
</tr>
<tr>
<td>Clear Sky</td>
<td>NA</td>
<td>-30.6</td>
<td>-30.1</td>
<td>-35.4</td>
</tr>
<tr>
<td></td>
<td>EU</td>
<td>-25.6</td>
<td>-24.6</td>
<td>-27.9</td>
</tr>
<tr>
<td></td>
<td>EA</td>
<td>-26.6</td>
<td>-26.3</td>
<td>-32.1</td>
</tr>
<tr>
<td></td>
<td>SA</td>
<td>-27.7</td>
<td>-27.8</td>
<td>-32.7</td>
</tr>
<tr>
<td></td>
<td>globe</td>
<td>-30.2</td>
<td>-29.9</td>
<td>-37.0</td>
</tr>
</tbody>
</table>
Table 4. Change of global annual mean AOD ($\tau \times 1000$, mean ± std. dev) in response to the 20% reduction of anthropogenic emissions in 4 source regions (SR6 - SR1) of 9 HTAP models.

<table>
<thead>
<tr>
<th>Source Region</th>
<th>$\tau_{SO4}$</th>
<th>$\tau_{POM}$</th>
<th>$\tau_{BC}$</th>
<th>$\tau_{SO4+POM+BC}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>NA</td>
<td>-0.67 ± 0.23</td>
<td>-0.054 ± 0.036</td>
<td>-0.026 ± 0.013</td>
<td>-0.75 ± 0.23</td>
</tr>
<tr>
<td>EU</td>
<td>-1.13 ± 0.47</td>
<td>-0.084 ± 0.058</td>
<td>-0.050 ± 0.021</td>
<td>-1.28 ± 0.45</td>
</tr>
<tr>
<td>EA</td>
<td>-0.97 ± 0.38</td>
<td>-0.142 ± 0.101</td>
<td>-0.086 ± 0.047</td>
<td>-1.19 ± 0.37</td>
</tr>
<tr>
<td>SA</td>
<td>-0.32 ± 0.11</td>
<td>-0.102 ± 0.059</td>
<td>-0.038 ± 0.019</td>
<td>-0.46 ± 0.10</td>
</tr>
<tr>
<td>Total</td>
<td>-3.10 ± 1.11</td>
<td>-0.383 ± 0.246</td>
<td>-0.199 ± 0.093</td>
<td>-3.70 ± 1.08</td>
</tr>
</tbody>
</table>
Table 5. Change of global annual mean aerosol DRF (unit: mWm\(^{-2}\), mean ± standard deviation, 1 mWm\(^{-2}\) = 0.001 Wm\(^{-2}\)) in response to the 20% reduction of anthropogenic emissions in 4 source regions as derived from analysis of SR1 and SR6 runs of 9 HTAP models.

<table>
<thead>
<tr>
<th>Source Region</th>
<th>DRF(_{SO4}) (mWm(^{-2}))</th>
<th>DRF(_{POM}) (mWm(^{-2}))</th>
<th>DRF(_{BC}) (mWm(^{-2}))</th>
<th>DRF(_{SO4+POM+BC}) (mWm(^{-2}))</th>
</tr>
</thead>
<tbody>
<tr>
<td>All-sky TOA DRF (mWm(^{-2}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>15.4 ± 5.1</td>
<td>1.4 ± 0.9</td>
<td>-3.0 ± 1.6</td>
<td>13.8 ± 5.3</td>
</tr>
<tr>
<td>EU</td>
<td>24.4 ± 8.9</td>
<td>1.6 ± 1.2</td>
<td>-5.9 ± 2.6</td>
<td>20.2 ± 9.5</td>
</tr>
<tr>
<td>EA</td>
<td>19.4 ± 7.9</td>
<td>3.1 ± 2.2</td>
<td>-9.4 ± 6.1</td>
<td>13.0 ± 9.9</td>
</tr>
<tr>
<td>SA</td>
<td>6.8 ± 1.9</td>
<td>2.5 ± 1.7</td>
<td>-3.6 ± 1.8</td>
<td>5.7 ± 2.6</td>
</tr>
<tr>
<td>Total</td>
<td>66.1 ± 22.8</td>
<td>8.6 ± 5.7</td>
<td>-21.9 ± 11.5</td>
<td>52.8 ± 25.3</td>
</tr>
<tr>
<td>All-sky surface DRF (mWm(^{-2}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>14.9 ± 4.9</td>
<td>1.9 ± 1.3</td>
<td>5.5 ± 2.4</td>
<td>22.4 ± 5.5</td>
</tr>
<tr>
<td>EU</td>
<td>23.0 ± 8.0</td>
<td>2.4 ± 1.7</td>
<td>8.8 ± 3.1</td>
<td>34.1 ± 7.3</td>
</tr>
<tr>
<td>EA</td>
<td>18.9 ± 7.5</td>
<td>4.4 ± 3.2</td>
<td>16.5 ± 8.7</td>
<td>39.9 ± 11.7</td>
</tr>
<tr>
<td>SA</td>
<td>6.9 ± 2.0</td>
<td>3.6 ± 2.3</td>
<td>8.1 ± 3.6</td>
<td>18.6 ± 4.6</td>
</tr>
<tr>
<td>Total</td>
<td>63.7 ± 21.4</td>
<td>12.3 ± 8.1</td>
<td>39.0 ± 16.6</td>
<td>115.0 ± 26.0</td>
</tr>
</tbody>
</table>
Table 6. Global annual average TOA all-sky forcing efficiency relative to emissions from the source regions (unit: mWm\(^{-2}\) per Tg) as derived from 8 HTAP models. The forcing efficiency for SO\(_4\) and POM is calculated with respect to SO\(_2\) and primary POM emissions, although some models with fully coupled chemistry include changes of SO\(_4\) resulting from the reductions of other emissions and a fraction of POM is secondary aerosol produced from a variety of volatile organic carbon.

<table>
<thead>
<tr>
<th>Source Region</th>
<th>SO(_4)</th>
<th>POM</th>
<th>BC</th>
</tr>
</thead>
<tbody>
<tr>
<td>NA</td>
<td>-3.6±1.0</td>
<td>-4.3±1.5</td>
<td>27.2±14.2</td>
</tr>
<tr>
<td>EU</td>
<td>-3.7±1.1</td>
<td>-4.2±1.6</td>
<td>36.7±18.0</td>
</tr>
<tr>
<td>EA</td>
<td>-2.7±0.9</td>
<td>-3.6±1.7</td>
<td>27.7±19.0</td>
</tr>
<tr>
<td>SA</td>
<td>-3.8±1.0</td>
<td>-4.0±1.7</td>
<td>25.0±13.6</td>
</tr>
<tr>
<td>Total</td>
<td>-3.3±0.9</td>
<td>-3.9±1.6</td>
<td>29.0±16.8</td>
</tr>
</tbody>
</table>
Table 7: Response of annual AOD and all-sky DRF (mean \pm std. dev.) in the receptor regions (columns) to 20% reductions of anthropogenic gas and aerosol emissions in the source regions (rows) estimated from 9 HTAP models. Percentage contributions of individual source regions are shown in parentheses for AOD (corresponding percentages for DRF are only slightly different and hence not shown). Contributions of a region on itself (termed “domestic”) are shown in bold. Here BC is listed separately from SO\(_4\) and POM combined (in which SO\(_4\) makes a dominant contribution of 69-94%, depending on region), because of their distinct effects on solar radiation. For SO\(_4\)+POM, surface DRF is slightly greater than TOA DRF because of weak absorption and not shown. A positive change of DRF for SO\(_4\)+POM suggests a weakened cooling effect by the emission reduction. A BC emission reduction introduces a negative change of TOA DRF (a weakened warming effect) and a positive change of surface DRF (a weakened cooling effect).

<table>
<thead>
<tr>
<th>Source Region</th>
<th>Receptor Region</th>
<th>NA</th>
<th>EU</th>
<th>EA</th>
<th>SA</th>
</tr>
</thead>
<tbody>
<tr>
<td>SO(_4)+POM AOD (x1000)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>-8.53±2.85 (82%)</td>
<td>-1.22±0.66 (7%)</td>
<td>-0.35±0.25 (2%)</td>
<td>-0.24±0.22 (2%)</td>
<td></td>
</tr>
<tr>
<td>EU</td>
<td>-0.42±0.21 (4%)</td>
<td>-13.81±4.43 (84%)</td>
<td>-1.21±0.68 (6%)</td>
<td>-2.18±0.79 (16%)</td>
<td></td>
</tr>
<tr>
<td>EA</td>
<td>-0.92±0.59 (9%)</td>
<td>-0.65±0.48 (4%)</td>
<td>-16.42±5.02 (84%)</td>
<td>-1.05±0.66 (8%)</td>
<td></td>
</tr>
<tr>
<td>SA</td>
<td>-0.54±0.91 (5%)</td>
<td>-0.72±1.32 (4%)</td>
<td>-1.68±0.95 (9%)</td>
<td>-10.37±3.21 (75%)</td>
<td></td>
</tr>
<tr>
<td>BC AOD (x1000)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>-0.28±0.09 (65%)</td>
<td>-0.05±0.04 (8%)</td>
<td>-0.02±0.01 (1%)</td>
<td>-0.01±0.01 (1%)</td>
<td></td>
</tr>
<tr>
<td>EU</td>
<td>-0.02±0.02 (4%)</td>
<td>-0.51±0.15 (76%)</td>
<td>-0.05±0.03 (4%)</td>
<td>-0.08±0.04 (8%)</td>
<td></td>
</tr>
<tr>
<td>EA</td>
<td>-0.10±0.09 (23%)</td>
<td>-0.07±0.08 (10%)</td>
<td>-1.04±0.28 (84%)</td>
<td>-0.09±0.06 (9%)</td>
<td></td>
</tr>
<tr>
<td>SA</td>
<td>-0.04±0.03 (8%)</td>
<td>-0.04±0.03 (6%)</td>
<td>-0.14±0.08 (11%)</td>
<td>-0.87±0.36 (82%)</td>
<td></td>
</tr>
<tr>
<td>SO(_4)+POM all-sky TOA DRF (mWm(^{-2}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>207.7±68.9</td>
<td>28.8±14.9</td>
<td>7.5±5.2</td>
<td>6.3±5.7</td>
<td></td>
</tr>
<tr>
<td>EU</td>
<td>10.4±5.42</td>
<td>304.4±93.3</td>
<td>25.4±14.1</td>
<td>54.2±20.3</td>
<td></td>
</tr>
<tr>
<td>EA</td>
<td>23.8±15.1</td>
<td>14.4±10.9</td>
<td>315.2±98.8</td>
<td>25.0±15.5</td>
<td></td>
</tr>
<tr>
<td>SA</td>
<td>11.2±15.5</td>
<td>8.8±9.4</td>
<td>32.9±15.4</td>
<td>256.4±80.4</td>
<td></td>
</tr>
<tr>
<td>BC all-sky TOA DRF (mWm(^{-2}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>-23.8±7.2</td>
<td>-5.1±3.3</td>
<td>-1.4±1.3</td>
<td>-1.5±1.3</td>
<td></td>
</tr>
<tr>
<td>EU</td>
<td>-1.7±1.4</td>
<td>-45.6±12.8</td>
<td>-4.5±2.7</td>
<td>-9.2±4.4</td>
<td></td>
</tr>
<tr>
<td>EA</td>
<td>-9.0±8.5</td>
<td>-6.7±7.3</td>
<td>-83.5±22.3</td>
<td>-8.6±6.5</td>
<td></td>
</tr>
<tr>
<td>SA</td>
<td>-3.0±2.3</td>
<td>-4.1±2.7</td>
<td>-11.8±5.3</td>
<td>-70.7±29.6</td>
<td></td>
</tr>
<tr>
<td>BC all-sky surface DRF (mWm(^{-2}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NA</td>
<td>64.1±20.5</td>
<td>11.5±7.2</td>
<td>3.3±3.0</td>
<td>3.5±2.8</td>
<td></td>
</tr>
<tr>
<td>EU</td>
<td>4.3±3.5</td>
<td>103.0±27.1</td>
<td>10.0±5.9</td>
<td>19.8±9.9</td>
<td></td>
</tr>
<tr>
<td>EA</td>
<td>22.7±21.6</td>
<td>14.4±15.5</td>
<td>212.8±57.0</td>
<td>22.3±15.0</td>
<td></td>
</tr>
<tr>
<td>SA</td>
<td>7.9±6.2</td>
<td>7.5±5.3</td>
<td>29.3±13.5</td>
<td>202.0±85.4</td>
<td></td>
</tr>
</tbody>
</table>
Figure 1: Illustration of four HTAP defined regions for examining the source-receptor relationships for anthropogenic aerosols: North America (NA, 15°-55°N, 60°W-125°W), Europe (EU, 25°-65°N, 10°W-50°E), East Asia (EA, 15°-50°N, 95°-160°E), and South Asia (SA, 5°-35°N, 50°-95°E). Regional and annual anthropogenic emissions of SO₂, primary POM, and BC from 8 models (excluding HADGEM2 and ECHAM5) are shown in bar charts, with error bar indicating the range of 8 models.
Figure 2: Annual average AOD for the external mixture of sulfate, POM, and BC simulated by baseline runs of 9 HTAP models.
Figure 3: Comparison of HTAP SR1 9-model median AOD for the external mixture of sulfate, POM, and BC (a) with MODIS-derived anthropogenic AOD over ocean as described in Yu et al. [2009] (b) and Bellouin et al. [2008] (c).
Figure 4: Comparisons of zonal variations of seasonal and 20°N-60°N average AOD for sulfate, POM, and BC combined as simulated by HTAP models (black line for median and shaded area for the range of 9 models) with the MODIS-derived over-ocean anthropogenic AOD (red line for Yu et al., 2009 and blue line for Bellouin et al., 2008). Note that HTAP model results cover both land and ocean, while MODIS-based anthropogenic AOD were estimated only over ocean.
Figure 5: Seasonal variations of 9-model average TOA DRF (a), and surface DRF (b) in all-sky conditions (Wm$^{-2}$) for the external mixture of sulfate, POM, and BC as derived from the HTAP baseline simulations (SR1).
Figure 6: Global annual mean forcing efficiency (unit: mWm$^{-2}$ per Tg) for sulfate (a) and BC (b) with respect to regional (denoted as NA, EU, EA, and SA) emissions as simulated by 8 HTAP models.
Figure 7: Annual average AOD (x100, top panel) and DRF (mWm$^{-2}$) at the top of atmosphere (TOA) (middle panel) and at the surface (bottom panel) in all-sky conditions resulting from 20% reductions of regional anthropogenic emissions over North America (NA), Europe (EU), East Asia (EA), and South Asia (SA), respectively. A positive value for DRF represents the reduced aerosol direct radiative forcing resulting from the reduction of emissions. Each of the four source regions is overlaid on corresponding SR6xx-SR1 maps where xx represents NA, EU, EA, and SA. Individual aerosol components are assumed to be mixed externally. The results are based on 9 HTAP models.
Figure 8: Seasonal variations of changes in the all-sky DRF (mWm\(^{-2}\)) at surface (SR6-SR1) due to 20% reduction of regional anthropogenic emissions derived from 9 HTAP models. Each of the four source regions is overlaid on corresponding SR6xx-SR1 maps where xx represents NA, EU, EA, and SA.
Figure 9: Ten-model (including ECHAM5) derived average relative annual intercontinental response (RAIR) for aerosol optical depth (AOD) in four receptor regions, by chemical component. Standard deviations, as indicated by error bars, reflect the model variability in simulating aerosol transport.
Figure 10: same as Figure 9 but for TOA all-sky DRF.