Radiative forcing due to major aerosol emitting sectors in China and India

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Understanding the radiative forcing caused by anthropogenic aerosol sources is essential for making effective emission control decisions to mitigate climate change. We examined the net direct plus indirect radiative forcing caused by carbonaceous aerosol and sulfur emissions in key sectors of China and India using the GISS-E2 chemistry-climate model. Diesel trucks and buses (67 mW m⁻²) and residential biofuel combustion (52 mW m⁻²) in India have the largest global mean, annual average forcings due mainly to the direct and indirect effects of BC. Emissions from these two sectors in China have near-zero net global forcings. Coal-fired power plants in both countries exert a negative forcing of about −30 mW m⁻² from production of sulfate. Aerosol forcings are largest locally, with direct forcings due to residential biofuel combustion of 580 mW m⁻² over India and 416 mW m⁻² over China, but they extend as far as North America, Europe, and the Arctic.


1. Introduction

Recent studies [e.g., Ramanathan and Carmichael, 2008; Shindell et al., 2012; Bond et al., 2013] have identified primary anthropogenic emissions of aerosols, particularly black carbon (BC) emissions from industrializing countries like China and India, as targets for emission control to mitigate climate change. The reasons are compelling: their contribution to radiative forcing is large and positive, the emission reductions would achieve immediate reductions in direct forcings, the emission reductions would be relatively easy and economical to achieve [UNEP/WMO, 2011], and major human health benefits would accrue locally. However, there has been limited study of the contributions to radiative forcing caused by the major source categories that release primary aerosols in China and India. The determination of net radiative forcing is complicated by the contributions from coemitted species and the complex physical and chemical interactions that occur among these primary emitted species and other components of the atmosphere. We examined emissions from the seven most important source categories in China and India and estimated the net radiative forcing from each of them, locally and around the globe.

2. Data and Methods

We used primary emissions of SO₂, BC, and OC for the year 2008 extracted from the 1996–2010 annual emission trends for China and India developed by Lu et al. [2011]. The emissions include all major anthropogenic sources as well as open biomass burning. Year-specific annual temporal distributions were developed for major sectors and gridded at a spatial resolution of 0.1° × 0.1°. Over the period 1996–2010, the emissions reveal generally upward trends with the exception of SO₂ emissions in China that decline after 2006. Annual emissions in the year...
selected for this study, 2008, are given in Figure 1. It should be noted that 2008 was the year of the Beijing Olympics; emissions were reduced somewhat from normal levels but only for a short time and in a small area, so this does not materially affect our results. Figures 5 and 6 of Lu et al. [2011] provide a complete comparison of the emission estimates with other studies, as well as estimates of uncertainty at 95% confidence. They show that the residential sector contributes the most to overall emission uncertainties in both China and India (60%–65% for BC and 67%–74% for OC).

For this analysis, we selected seven cases representing those sectors in China and India that contribute the most BC, OC, and SO2. For BC, the sectors are diesel trucks and buses in the transportation sector (6.4% of total BC emissions in China and 4.4% in India), residential biofuel combustion (34.5% in China and 49.8% in India), and residential fossil fuel combustion (20.6% in China and very low in India). For OC, residential biofuel combustion (60.4% in China and 71.7% in India) and residential fossil fuel combustion (6.2% in China, very low in India) are important. For SO2, we selected fossil fuel combustion in the power sector, which contributes 32.2% of total SO2 emissions in China and 52.3% in India, together with residential fossil fuel combustion in China, which contributes 8.7%. Figure 1 shows the spatial distributions of emissions of BC, OC, and SO2 in China and India and the relative contributions of key sectors.

These year 2008 emissions are used with the NASA GISS chemistry-climate model GISS-E2. This is the latest version of the GISS model updated from Schmidt et al. [2006] and including gaseous and aerosol chemistry as described in Shindell et al. [2013a] and references therein. All emissions other than those described above for BC, OC, and SO2 in China and India are based on year 2000 of the historical transient emissions used in the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Lamarque et al., 2010] and are held constant. Background conditions such as greenhouse-gas levels are for the year 2000, while ocean conditions are fixed at observed 1996–2004 averages. Simulations were performed by removing BC, OC, and SO2 emissions from one region and sector at a time. We analyzed the last 32 years of 35-year simulations, examining the difference relative to a control simulation in all cases. The long integrations are required to allow diagnoses of indirect aerosol effects. We calculated the total effective radiative forcing (ERF), defined as the net top-of-atmosphere downward flux change after allowing for all responses other than those of the ocean (sea-surface temperatures and sea-ice cover are fixed). This forcing includes rapid adjustments in the atmosphere in response to aerosols, especially cloud responses, and has been shown to provide a better indication of the eventual climate response than the traditional radiative forcing that includes adjustment of stratospheric temperature only [Hansen et al., 2005; Lohmann et al., 2010]. ERF includes the response of clouds, water vapor, lapse rate, etc., as well as aerosol direct effects, and hence encompasses the indirect effects referred to as semidirect, cloud albedo, and cloud lifetime [Shindell et al., 2013b]. We hereafter refer to ERF minus the direct forcing as the indirect forcing. In comparison with AERONET observations, the GISS-E2 model underestimates both aerosol optical depth and absorbing aerosol optical depth over East and South Asia by ~30%–50%, consistent with nearly all global models [Shindell et al., 2013b]. These low biases suggest that the values reported here may be conservative.

Figure 1. Emissions distributions in 2008 for China and India at 0.1° × 0.1° spatial resolution for (top) BC, (middle) OC, and (bottom) SO2, also showing totals for each species and shares for each source category.
The effect of emissions from residential biofuel combustion is small, the net emissions contributions are not the same, and the changes. Surface forcing due to ozone is not important and not calculated. Indirect forcing uncertainty is 35 mW m\(^{-2}\). Values in parentheses are the corresponding radiative forcing estimates at the surface in the same units.

### Results

Table 1 presents global mean, annual average radiative forcings due to all aerosols and ozone resulting from the indicated sector in the given region. Biofuel combustion in the residential sectors of both India and China shows a significant positive direct radiative forcing from BC emissions of about 50 mW m\(^{-2}\), roughly 5% of which is due to albedo effects. These effects are only slightly offset (~15%) by the negative forcing due to coemitted OC, which is <10 mW m\(^{-2}\).

For coal-fired power, BC is offset by large indirect effects from residential emissions for China, whereas indirect forcing is strongly positive for the diesel trucks and buses case in India. This positive indirect forcing suggests that in that instance, the semidirect impact of BC is larger than the microphysical aerosol-cloud impacts of BC and OC in this model, which was also seen in previous work [Shindell et al., 2012]. Not only do different climate regimes influence the country comparisons, but the emissions contributions are not the same, which affects the balance of BC and OC emissions. With other effects from residential biofuel emissions being small, the net effect of emissions from residential biofuel combustion is 52 mW m\(^{-2}\) for India and near zero for China. Increasing the albedo forcing of BC by a factor of 3–5 to account for the stronger surface temperature response per unit forcing for this agent relative to others [Flanner et al., 2007; Koch et al., 2009] adds ~10 mW m\(^{-2}\) to the net forcing.

For coal-fired power plants in both India and China, we found negligible effects from carbonaceous aerosols due to their very low emissions, while the direct radiative effects of sulfate (positive) and nitrate (negative) roughly offset each other. Thus, negative indirect forcings of ~18 mW m\(^{-2}\) in India and ~30 mW m\(^{-2}\) in China largely drive the net forcings of ~31 mW m\(^{-2}\) in India and ~33 mW m\(^{-2}\) in China from this source type. An additional simulation for coal-fired power plants in China in which only SO\(_2\) emissions were changed shows that it is the sulfate associated with high SO\(_2\) emissions that contributes virtually all of the net forcing (Table 1). This simulation also demonstrates that the nitrate forcing is a response to the SO\(_2\) changes.

For the transportation sector, we examined the combined effect of diesel trucks and buses. The results for India and China are significantly different. For India, we found that all direct forcings are small, and it is the large indirect forcing that determines the net positive radiative forcing of 67 mW m\(^{-2}\). In the case of China, the direct effect of BC is more pronounced, while the indirect forcing is comparable in magnitude but of the opposite sign, yielding a total net forcing of only 2 mW m\(^{-2}\). The positive cloud forcing in response to diesel emissions in India indicates a strong positive indirect BC forcing, in contrast to the negative indirect forcing associated with sulfur-rich, coal-fired power-plant emissions (consistent with our knowledge of the cloud response to absorbing versus reflective aerosols). Interestingly, the magnitude of the positive indirect and BC direct forcings are comparable for the BC-rich and low-sulfur diesel transportation sector emissions from China, while for India the indirect forcing is much larger, indicating that this response is highly dependent upon conditions such as the regional aerosol residence times and/or background aerosol loading. Similarly, the indirect effect in response to residential biofuel emissions is substantially more negative for China than for India despite similar OC direct forcing. Given the positive indirect forcing in response to Indian diesel emissions, this suggests that the negative indirect effect of OC is greater relative to the indirect effect of BC for China than for India. Hence, indirect responses (primarily clouds) appear to be more positive for...
emissions from India than from China, perhaps because of the large difference in the surface albedo of the downwind Himalayas and Pacific Ocean, respectively.

While the impacts of coemitted gaseous pollutants such as NOx and CO are not considered here, they have been evaluated in prior studies with a similar GISS model [Shindell et al., 2008]. In that work, radiative forcing due to ozone changes was −15% of the sulfate direct forcing for the power sector and 25% and 31% of the BC forcing for the surface transportation and residential sectors, respectively. Adjusting the ozone responses found here to match those results would change the ozone forcing by <5 mW m⁻² for the power and transportation sectors, while adding 17 mW m⁻² to the residential sector in India and 7 mW m⁻² to both the residential subsectors in China. These values are well within the uncertainty in the aerosol indirect forcing. Global forcing at the surface is also presented in Table 1. Only the three residential cases have significant direct forcings, negative in each case due to BC. There is also a substantial reduction in surface forcing due to the indirect effect for most sectors, which may be related to the semidirect effect of BC. Biofuel combustion typically has the largest effect, with comparable forcing from coal-fired power generation in India.

Figure 2 presents the spatial distribution of the direct plus BC-albedo radiative forcings for each of the seven cases. The indirect forcing values are typically not statistically significant at the local scale, so are excluded in this analysis. As expected, forcing is concentrated locally, where the emissions are highest. For China, however, effects extend throughout the Northern Hemisphere, particularly for the residential cases, primarily owing to long-range transport of BC. For India, maximum forcing is generally collocated with emissions, though Southwest China and the Himalayas also experience high values. For the cases of coal-fired power plants in both India and China, the forcing is significantly negative locally and in the global mean but contains major areas of positive forcing, which can be attributed to removal of ammonium during sulfate formation leading to reduced nitrate formation away from the source region. There are also positive forcings in cryosphere regions, for example, the Himalayas in the case of Indian power plants, which are attributable to BC direct and albedo forcing.

Table 2 presents the direct radiative forcing estimates averaged over 10 world regions, as well as China and India themselves. The aerosol forcings are largest locally, with direct forcings as high as 580 mW m⁻² averaged over India.
for the residential biofuel case and 416 mW m\(^{-2}\) averaged over China also for the residential biofuel case. In addition, however, aerosol transport leads to significant direct forcings in remote regions, particularly due to emissions from Chinese sources, which are as high as 82 mW m\(^{-2}\) averaged over North America, 94 mW m\(^{-2}\) averaged over Europe, and 147 mW m\(^{-2}\) in the Arctic. The spatial distribution of surface forcing is similar to the direct forcing shown in Figure 2.

[18] There is substantial uncertainty in radiative forcing by short-lived species, especially for aerosols. Two recent assessments have examined the forcing due to BC and OC in particular, as well as sulfate [UNEP/WMO, 2011; Bond et al., 2013]. Bond et al. [2013] provide a best estimate of the historical BC direct forcing that is larger than in the GISS model, but they also attribute a large negative indirect forcing to OC. UNEP/WMO [2011] also provide a best estimate of the historical BC direct forcing that is larger than that in the GISS model but attribute a weaker indirect effect to BC. Scaling our results to match either study’s direct forcing estimates and indirect/direct forcing ratios (as in UNEP/WMO [2011]) causes the net forcing from Indian or Chinese residential biofuel emissions to increase to \(\sim 100\) mW m\(^{-2}\), roughly doubling the Indian value. For Indian diesel, while the scaled value using Bond et al. [2013] also produces a positive indirect forcing, due to the dominance of BC’s indirect effects over OC’s (as in our modeling), its value is weaker and the UNEP scaling gives near zero indirect, leading to substantially lower negative indirect effect of sulfate while the indirect effect for Chinese power-plant emissions is quite similar, however, the near-zero result found here. Forcing due to Indian and Chinese power-plant emissions is quite similar, however, unsurprisingly as these results almost exclusively from the negative indirect effect of sulfate while the indirect effect for other sectors is the result of offsetting positive and negative values due to BC and OC, respectively. Thus, the large positive impacts estimated here for emissions from Indian biofuels are robust; for Chinese transportation and residential biofuel, they appear to be conservative, while for Indian transportation, they may be on the high side, based on the values in the GISS model relative to these other two studies. Note that the scaling is based only on global mean forcings and does not account for the large regional variations found here for indirect effects. As such, the internally calculated results may be more accurate, but analysis of additional models is required to confirm those results.

[16] The radiative forcing effects caused by emissions from these seven sector/region combinations gives some indication of which of them should be preferential targets for emissions control under a climate mitigation policy. However, a direct intercomparison can be misleading because there are widely differing amounts of energy generated and fuels consumed in the seven cases. It is useful to normalize the forcings according to the amount of energy associated with each case. Normalizing to the energy amounts reported by Lu et al. [2011], we calculated total forcings (from Table 1) per exajoule of associated energy to be dominated by Indian diesel trucks and buses at 81 mW m\(^{-2}\) EJ\(^{-1}\), followed by Indian residential biofuel at 6.6 mW m\(^{-2}\) EJ\(^{-1}\). These sectors could be considered as preferred targets for mitigation actions on the basis of their contributions to forcing; of course, other attributes such as cost and feasibility need to be taken into account, as well as the ancillary impacts on human health, but nevertheless, these sectors match well with mitigation targets recommended in other work [e.g., Shindell et al., 2012].

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