

Estimating ground-level PM_{2.5} using aerosol optical depth determined from satellite remote sensing

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[1] We assess the relationship of ground-level fine particulate matter $(PM_{2.5})$ concentrations for 2000-2001 measured as part of the Canadian National Air Pollution Surveillance (NAPS) network and the U.S. Air Quality System (AQS), versus remotesensed PM_{2.5} determined from aerosol optical depths (AOD) measured by the Moderate Resolution Imaging Spectroradiometer (MODIS) and the Multiangle Imaging Spectroradiometer (MISR) satellite instruments. A global chemical transport model (GEOS-CHEM) is used to simulate the factors affecting the relation between AOD and $PM_{2.5}$. AERONET AOD is used to evaluate the method (r = 0.71, N = 48, slope = 0.69). We find significant spatial variation of the annual mean ground-based measurements with $PM_{2.5}$ determined from MODIS (r = 0.69, N = 199, slope = 0.82) and MISR (r = 0.58, N = 199, slope = 0.57). Excluding California significantly increases the respective slopes and correlations. The relative vertical profile of aerosol extinction is the most important factor affecting the spatial relationship between satellite and surface measurements of $PM_{2,5}$; neglecting this parameter would reduce the spatial correlation to 0.36. In contrast, temporal variation in AOD is the most influential parameter affecting the temporal relationship between satellite and surface measurements of PM_{25} ; neglecting daily variation in this parameter would decrease the correlation in eastern North America from 0.5-0.8 to less than 0.2. Other simulated aerosol properties, such as effective radius and extinction efficiency have a minor role temporally, but do influence the spatial correlation. Global mapping of PM2.5 from both MODIS and MISR reveals annual mean concentrations of 40-50 ug/m³ over northern India and China.

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1. Introduction

[2] Exposure to fine particulate matter with diameter less than 2.5 um (PM_{2.5}) has numerous negative effects upon human health, including cancers of the lung, pulmonary inflammation and cardiopulmonary mortality [*Atkinson et al.*, 2001; *Pope et al.*, 2002]. Global measurements of these aerosols would be valuable to epidemiological studies, the design of air quality control strategies, and air quality forecasting [*Al-Saadi et al.*, 2005].

[3] The launch of the Moderate Resolution Imaging Spectroradiometer (MODIS) and the Multiangle Imaging Spectroradiometer (MISR) instruments onboard NASA's Terra satellite in 1999 has provided global measurements of aerosol optical depth (AOD), a measure of light extinc-

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tion by aerosol in the atmospheric column, during their overpass time of 1030 local time (LT). The temporal correlation between space-based measurements of AOD and surface PM concentrations has received considerable attention [i.e., Engel-Cox et al., 2004a, 2004b, 2005; Chu et al., 2003]. The quality of the correlation varies greatly with region, generally being much higher over the eastern United States as compared to the western United States [Al-Saadi et al., 2005]. Empirical relationships between remotely measured AOD and surface PM_{2.5} have been developed for the southeast United States using both MODIS [Wang and Christopher, 2003] and MISR [Liu et al., 2005]. A common concern is the dependence upon several factors in addition to the AOD measurement, including the aerosol vertical profile, aerosol type and atmospheric conditions. Liu et al. [2004a] developed a simple, yet effective approach to correct for spatial and seasonal variation in these factors by applying local scaling factors from a global atmospheric chemistry model to AOD retrieved from MISR:

$$Estimated PM_{2.5} = \frac{Model \ surface \ aerosol \ concentration}{Model \ AOD} \times Retrieved \ AOD \tag{1}$$

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Here we investigate the relationship between satellitemeasured AOD and surface $PM_{2.5}$ at satellite overpass time. Section 2 presents an explicit formulation of the factors involved and our approach to determine surface $PM_{2.5}$. In section 3 we compare ground-level $PM_{2.5}$ against concentrations estimated from both MODIS and MISR AOD. We also validate this approach using AERONET AOD to estimate $PM_{2.5}$. We then assess temporal and spatial variability of the parameters in that relationship to determine which factors deserve the most attention to improve $PM_{2.5}$ estimates derived from satellite measurements of AOD. Finally, we provide conclusions regarding the most influential parameters on the accuracy of this technique.

2. Determination of PM_{2.5} From Satellite Measurements of AOD

[4] We first develop an explicit formulation of the relationship between AOD and $PM_{2.5}$ in order to isolate the parameters involved. Then we describe our simulation of that relationship.

2.1. AOD-PM Relation

[5] AOD, τ , and total column aerosol mass loading Ω are related by:

$$\Omega = \frac{4}{3} \frac{\rho r_{eff} \tau}{Q_e} \tag{2}$$

where ρ is the aerosol mass density at ambient relative humidity, r_{eff} is the column averaged effective radius (defined as the ratio of the third to second moment of an aerosol size distribution at ambient relative humidity), and Q_e is the column averaged extinction efficiency.

[6] Similarly, by accounting for the relative vertical profile of aerosol extinction, the aerosol mass concentration $M_{\Delta z}$ between the ground and altitude Δz can be expressed as

$$M_{\Delta z} = \begin{bmatrix} \frac{4}{3} \frac{\rho_{\Delta z} r_{\Delta z, eff} f_{\Delta z}}{Q_{\Delta z, e} \Delta z} \end{bmatrix} \tau \tag{3}$$

where f represents the fractional optical thickness below altitude Δz . All parameters in the bracketed expression refer to representative values below altitude Δz .

[7] Surface $PM_{2.5}$ measurements are usually of dry aerosol mass as described in section 3. Assuming spherical aerosols and accounting for aerosol hygroscopicity, the dry mass of $PM_{2.5}$ at the surface can expressed as:

$$M_{2.5,d,\Delta z} = \left[\frac{4}{3} \left(\frac{r_{2.5,d,\Delta z,eff}}{r_{2.5,\Delta z,eff}}\right)^3 \left(\frac{\rho_{2.5,d,\Delta z} r_{2.5,\Delta z,eff} f_{2.5,\Delta z}}{Q_{2.5,e,\Delta z} \Delta z}\right)\right] \tau \quad (4)$$

where the subscript *d* indicates dry conditions and the subscript 2.5 denotes aerosols smaller than 2.5 um in diameter. $M_{2.5,d,\Delta z}$ is the total fine dry aerosol mass between the surface and altitude Δz , $r_{2.5,d,\Delta z,eff}$ is the fine dry effective radius, and $f_{2.5,\Delta z}$ is the ratio of fine AOD below altitude Δz to the total AOD. This equation assumes uniform aerosol properties between the surface and altitude

 Δz . We refer to $M_{2.5,d,\Delta z}$ as remote-sensed PM_{2.5}. Equation (4) reduces to equation (1) if a model is used to provide the values within the brackets.

2.2. Simulation of Parameters Affecting AOD-PM Relation

[8] We use a chemical transport model (GEOS-CHEM) to calculate the values within brackets in equation (4). The GEOS-CHEM chemical transport model (http://wwwas.harvard.edu/chemistry/trop/geos/index.html) is driven by assimilated meteorological data from the Goddard Earth Observing System (GEOS-3) at the NASA Global Modeling Assimilation Office (GMAO). Meteorological fields include surface properties, humidity, temperature, winds, cloud properties, heat flux and precipitation. We use GEOS-CHEM v7-02-01 with 30 levels and 50 tracers using a $1^{\circ} \times 1^{\circ}$ nested model version for North America and the adjacent oceans (10-60°N, 40-140°W) with dynamic boundary conditions from a global $4^{\circ} \times 5^{\circ}$ simulation. A detailed description of this one-way nesting in the GEOS-CHEM model is given by Wang et al. [2004]. The lowest five levels are centered at approximately 10, 50, 100, 200 and 300 meters; a typical value for Δz , the lowest model level in equation (4), is approximately 20 m, varying with surface pressure. We assume that PM2.5 concentration does not vary substantially from the surface to this height.

[9] The GEOS-CHEM aerosol simulation includes the sulfate-nitrate-ammonium system, carbonaceous aerosols, mineral dust, and sea salt. The aerosol and oxidant simulations are coupled through formation of sulfate and nitrate [Park et al., 2004], heterogeneous chemistry [Jacob, 2000] and aerosol effects of photolysis rates [Martin et al., 2003]. Wet and dry deposition are based upon Liu et al. [2001], including both washout and rainout of hydrophilic carbonaceous species. Primary organic aerosol emissions are assumed to be 50% hydrophilic, while secondary organic aerosols are assumed to have a wet scavenging efficiency of 80% [Park et al., 2003]. We use seasonal average biomass burning inventories based upon climatological emissions. Anthropogenic carbon emission estimates are based upon Cooke et al. [1999] over North America, with imposed seasonal variation from Park et al. [2003]. Secondary organic chemistry is based upon the work by Chung and Seinfeld [2002]. The dust simulation is from T. D. Fairlie et al. (The impact of transpacific transport of mineral dust in the United States, submitted to Atmospheric Environment, 2006).

[10] Previous evaluations of the aerosol simulation have found a high degree of consistency with observations. Annual mean simulated sulfate concentrations explain 75–90% of the spatial variance in surface measurements from the IMPROVE network with little bias; slight underestimates are found versus IMPROVE for organic and elemental carbon ($r^2 = 0.67$, slope = 0.74; $r^2 = 0.84$, slope = 0.85, respectively) [*Park et al.*, 2004]. The simulated vertical variation in aerosol extinction is typically within 25% of lidar observations at the DOE/ARM site in Oklahoma and Cheju Island in South Korea [*Hu et al.*, 2006].

[11] We conduct a near 2 year simulation over January 2001 to October 2002. The GEOS-3 meteorological fields used for our simulation are not available after October 2002.



Figure 1. Seasonal average of surface PM2.5 concentrations for December–February (DJF), March–May (MAM), June–August (JJA), September–November (SON) during the period from January 2001 to October 2002. (top) Ground-level measurements from the combined NAPS/AQS network. (middle and bottom) Determined respectively from MODIS and MISR measurements of aerosol optical depth (AOD). AOD measurements above 1.5 were discarded to reduce cloud contamination. White areas indicate ocean or regions with AOD measurements on fewer than either 40% of days for MODIS or 8% of days for MISR.

Local values for each parameter in equation (4) are taken from the simulation for each day at the MODIS and MISR overpass times between 1000 and 1200 local time. The parameters in equation (4) for each aerosol type are treated individually as an external mixture.

3. Measurements of Surface PM_{2.5}

[12] Here we compare ground-based measurements of surface $PM_{2.5}$ with remote-sensed concentrations determined from AOD using the relationship in section 2.

3.1. Ground-Based Surface PM_{2.5}

[13] The ground-based measurements are from Environment Canada's National Air Pollution Surveillance (NAPS) Network and the U.S. Environmental Protection Agency's Air Quality System (AQS). Both networks collect continuous measurements primarily from tapered element oscillating microbalance (TEOM) instruments, which infer the collected PM_{2.5} mass from changes in the natural frequency of the oscillator. Samples are heated to 30° – 50° C to ensure dry conditions.

[14] Figure 1 (top) shows seasonal mean $PM_{2.5}$ concentrations taken on the hour, between 1000 and 1200 LT inclusive, as measured by the combined NAPS/AQS network from January 2001 to October 2002. $PM_{2.5}$ in the northeast United States exhibits a seasonal maximum during summer, corresponding to higher sulfate levels produced by increased SO₂ oxidation rates [*Chin et al.*, 2000]. Elevated $PM_{2.5}$ in California is associated with high emissions of aerosol precursors and topography that is exacerbated in winter by low mixing height, coupled with the thermody-

namic tendency for ammonium nitrate to exist in aerosol phase at lower wintertime temperatures [*Blanchard and Tanenbaum*, 2003]. Southeastern PM_{2.5} concentrations are driven largely by organics, emitted directly by fires and formed as secondary organic aerosol from biogenic hydrocarbons; processes which are most active during warm seasons [*Malm et al.*, 2004].

3.2. Surface PM_{2.5} Estimated From MODIS AOD

[15] The MODIS aerosol retrieval is summarized by Remer et al. [2005]. Seven channels are used: 0.47, 0.55 0.66, 0.87, 1.24, 1.64 and 2.1 um. MODIS provides near global coverage on a daily basis. Radiation exiting the atmosphere can be approximated as a function of optical depth, scattering phase function and single-scattering albedo when over a dark surface [Kaufman et al., 1997]. Over land, the MODIS aerosol algorithm exploits this relation to measure optical depth by locating dark surfaces with the 2.1 um channel that is transparent to fine-mode particles and by empirically relating the 2.1 um measurements to surface properties at visible wavelengths. A simple model provides additional information on single scattering albedo and scattering phase function, with look-up tables used to determine aerosol properties at 0.47 and 0.66 um. Over water, spatial variability in the 0.55 um channel, high reflectance in the 0.47 um channel and infrared channels are all used to locate and avoid pixels contaminated with clouds. The 1.38 um channel is used to locate cirrus clouds. Chu et al. [2003] found significant agreement between MODIS and AERO-NET measurements of AOD (r = 0.82 - 0.91, slope = 0.83). MODIS retrievals of AOD over desert and coastal sites are, however, biased high because of errors caused by surface



Figure 2. Comparison of average surface $PM_{2.5}$ from January 2001 to October 2002 determined from ground measurements versus surface $PM_{2.5}$ inferred from MODIS and MISR measurements of aerosol optical depth. NAPS/AQS averages are compiled between 1000 and 1200 LT, during successful overpass measurements. Eastern (crosses) and western (circles) sites are separated at 96°W. Measurements of the California region are indicated by circled stars. The solid line represents y = x. The dashed line was calculated with reduced major axis linear regression [*Hirsch and Gilroy*, 1984].

brightness and subpixel water contamination [Abdou et al., 2005].

[16] Figure 1 (middle) shows seasonally averaged $PM_{2.5}$ concentrations calculated using $1^{\circ} \times 1^{\circ}$, daily level-3 version 4 MODIS AOD in equation (4). MODIS-estimated $PM_{2.5}$ generally compares well with the surface measurements, capturing both a similar structure and magnitude. Both show a seasonal maximum over the eastern United States during summer and low values in the Midwest throughout the year. However, surprisingly high values of surface $PM_{2.5}$ are found in the southwestern United States where few $PM_{2.5}$ measurements exist from AQS. Measurements from the Interagency Monitoring of Protected Visual Environments (IMPROVE) network indicate much lower

 $PM_{2.5}$ concentrations in this region than we observe with MODIS [*Malm et al.*, 2004], suggesting an artifact in the retrieval. $PM_{2.5}$ determined from MODIS typically overpredicts surface measurements by 3-5 ug/m³ at coastal sites on the Atlantic ocean and the Gulf of Mexico.

[17] Figure 2 (top) compares annual averages of coincident daily MODIS PM_{2.5} and surface PM_{2.5} for January 2001 to October 2002. A significant correlation (r = 0.69, N = 199) with a slope of 0.82 is found between MODIS PM_{2.5} and surface PM_{2.5}. However, MODIS PM_{2.5} is biased high by 5.1 ug/m³ on average.

3.3. Surface PM_{2.5} Estimated From MISR AOD

[18] The MISR retrieval algorithm is summarized by Martonchik et al. [2002]. MISR is fitted with nine cameras having aft and forward viewing angles of 70.5°, 60.0°, 45.6°, 26.1° and nadir viewing at four spectral bands of 0.446, 0.558, 0.672 and 0.866 um. The MISR instrument requires between 6 and 9 days for complete global coverage because of its smaller across track viewing angle compared to MODIS. The MISR algorithm infers the surface-leaving and path radiance contributions to total observed radiance without any assumption regarding the absolute surface reflectance by observing radiometric changes leaving the top of the atmosphere from a single location but at different viewing angles. The surface-leaving radiance is then compared against precalculated values contained in look-up tables to determine the best fitting aerosol composition and associated AOD. MISR AOD and AERONET are generally consistent, with correlation coefficients ranging from ~ 0.7 to 0.9 depending upon region [Kahn et al., 2005; Liu et al., 2004b].

[19] Figure 1 (bottom) shows remote-sensed $PM_{2.5}$ calculated using daily level-3 MISR AOD in equation (4). Locations having less than 8% of days during the period of study represented were removed. This threshold is necessarily lower than MODIS owing to MISR's 6-9 day global coverage. MISR PM_{2.5} is generally consistent with surface measurements during summer, however, more scatter is evident in winter. MISR PM2.5 concentrations in the southwest are only 10-20% of MODIS PM_{2.5} concentrations, and more consistent with GEOS-CHEM simulations. Concentrations in California remain biased low versus NAPS/ AQS measurements. Accurate measurements of AOD are expected over bright surfaces such as deserts [Martonchik et al., 2004]. The bias may result from a regional bias in GEOS-CHEM ammonium nitrate [Park et al., 2006]. An underestimate of surface emissions in GEOS-CHEM would influence the vertical structure in equation (4), lowering remote-measured PM_{2.5} concentrations.

[20] Figure 2 (bottom) compares MISR $PM_{2.5}$ against NAPS/AQS measurements. The correlation between MISR $PM_{2.5}$ and NAPS/AQS $PM_{2.5}$ is significant (r = 0.58, N = 199), with a slope of 0.57 and a mean positive bias of 3.1 ug/m³. The correlation and slope found here for MISR are lower than that given by *Liu et al.* [2004a] where California was excluded from the comparison. The removal of California from comparison improves the comparison with MISR $PM_{2.5}$ (r = 0.61, N = 189, slope = 0.68) with little effect versus MODIS $PM_{2.5}$ (r = 0.67, N = 189, slope = 0.93). Another major difference between these two studies is the spatial resolution of the analysis. Maintaining the removal of California while degrading the



Figure 3. Seasonally averaged AERONET-derived $PM_{2.5}$ concentrations January 2001 to October 2002. AOD were taken between 1000 and 1200 LT and required to have measurements on at least 8% of days per season.

resolution of MISR PM_{2.5} to 2° by 2° further improves the correlation (r = 0.69, slope = 0.70) with surface measurements. Similarly degrading MODIS PM_{2.5} results in little change (r = 0.69, slope = 0.86). Model developments from GEOS-CHEM v5-07-08 used by *Liu et al.* [2004a] to v7-01-02 used here also contributes to differences in the relationship between MISR PM_{2.5} and surface values.

3.4. Surface PM_{2.5} Estimated From AERONET AOD

[21] The Aerosol Robotic Network (AERONET) is a globally distributed network of CIMEL spectral radiometers operating at seven spectral bands (0.34, 0.38, 0.44, 0.50, 0.67, 0.87 and 1.02 um). A detailed description of these automatic-tracking, sun and diffuse sky radiometers is given by *Holben et al.* [1998]. AERONET measurements of AOD have an uncertainty of $\sim 0.01-0.02$ [*Holben et al.*, 2001]. Determination of surface PM_{2.5} from AERONET could provide a test of our method by reducing uncertainties associated with satellite-based AOD measurements.

[22] Figure 3 shows $PM_{2.5}$ determined from level 2.0 AERONET AOD using equation (4). We find a significant seasonal and spatial relationship with ground-based $PM_{2.5}$ (r = 0.71, N = 48, slope = 0.69), biased low by an average of 2.5 ug/m³. If California is excluded from the statistics the relationship improves to r = 0.80, N = 43, slope = 0.89. AERONET PM_{2.5} shows low concentrations throughout the southwest, in contrast with MODIS PM_{2.5}, providing evidence of a regional bias in MODIS AOD, since other parameters from equation (4) have remained constant irregardless of AOD source. This is consistent with *Abdou*

et al. [2005], who found that MODIS AOD is biased against AERONET in desert regions.

[23] $PM_{2.5}$ determined from AERONET AOD in California is substantially underpredicted, as was found for MISR $PM_{2.5}$. This regional bias may reflect a weak gradient in the model extinction profile, as noted earlier. As a result, the apparent superior performance of MODIS $PM_{2.5}$ over MISR $PM_{2.5}$ in California may result from a positive bias in MODIS AOD combined with an unrealistic vertical structure that underestimates the fraction of AOD near the surface.

4. Factors Affecting the Relationship Between Retrieved and Measured Surface PM_{2.5}

[24] An advantage of using equation (4) for $PM_{2.5}$ prediction is that it allows the isolation of individual variables and thereby the assessment of which parameters must be represented most accurately. Here we investigate each variable for both spatial and temporal variation over the time period of this study.

[25] Table 1 summarizes the most important factors affecting the spatial correlation of mean surface and remotely measured $PM_{2.5}$ as determined by replacing each parameter with a representative constant. Spatial variation in the relative vertical profile of modeled aerosol extinction has the largest effect on the accuracy of both mean remote $PM_{2.5}$ measurements; neglect of this parameter in MODIS and MISR reduces the spatial correlation versus surface measurements to 0.36 and 0.37, respectively. The AOD itself also exhibits significant influence on the accuracy of

Table 1. Spatial Correlation Coefficient, r, Between Remote $PM_{2.5}$ and Surface Measurements

	MODIS	MISR
Standard	0.69	0.58
Constant vertical structure	0.36	0.37
$(f_{2.5,\Delta z})$		
Constant AOD	0.58	0.46
Constant aerosol properties	0.61	0.40
$(Q_{2.5,e}, r_{2.5,\Delta z,eff}, r_{2.5,d,\Delta z,eff}, \rho_{2.5,d,\Delta z})$		

remotely measured surface concentrations with respect to surface measurements. The spatial distribution of other aerosol properties is of little importance, except in the California region, which drives a significant change in the correlation when aerosol properties are held constant.

[26] Figure 4 shows the temporal relationship between remote and surface measurements of $PM_{2.5}$ under the same conditions as Table 1. Strong temporal correlation with



Figure 4. Temporal correlation between daily remote and surface measurements of $PM_{2.5}$ between January 2001 and October 2002 for a standard case and three sensitivity studies in which parameters in equation (4) are temporally invariant. Correlations have minimum AOD measurements on either 40% of days for MODIS or 8% of days for MISR.



Figure 5. Time series plots of mean ground-level $PM_{2.5}$ between 1000 and 1200 LT for Toronto, Ontario, Galveston, Texas, Fayetteville, North Carolina, and Kent, Washington. MODIS derived $PM_{2.5}$ is plotted in blue, MISR $PM_{2.5}$ is plotted in red, and NAPS/AQS $PM_{2.5}$ is plotted in black.

surface measurements exist in the east (r = 0.5–0.8) and a poor correlation in the west (r < 0.3). Replacing the simulated vertical structure with a constant vertical structure shows that the simulated vertical structure improves slightly the temporal correlation in California, but also slightly decreases agreement in the east. The quality of the temporal correlation is determined largely by the temporal variation in AOD; exclusion of this parameter removes almost all temporal agreements, with the exception of California for which a moderate correlation (r ~ 0.4) remains. The temporal variation of other parameters has an insignificant effect on the relationship between AOD and surface PM_{2.5}.

[27] Figure 5 compares time series of surface $PM_{2.5}$ at four sites and the satellite retrieval. Overall correlations at these sites for MODIS and MISR, respectively, are: Toronto, Ontario: r = 0.67, r = 0.35; Galveston, Texas: r = 0.34, r = 0.48; Fayetteville, North Carolina: r = 0.53, r = 0.48; and Kent, Washington: r = 0.09, r = -0.11. MODIS $PM_{2.5}$ measurements in Kent show a distinct loss in accuracy during June–October, remaining much closer to NAPS/

AQS PM_{2.5} during other times of the year. As a result, Kent's overall correlation is quite low, typical of the northwestern United States as shown in Figure 4. MODIS cannot capture AOD measurements over Toronto during winter because of the presence of snow, coinciding with a time in which MISR PM_{2.5} shows a loss of consistency with surface measurements. The Toronto region exhibits one of the strongest correlations between remote PM_{2.5} and NAPS/ AQS PM_{2.5}. *Wang and Christopher* [2003] found similarly high correlations in Jefferson County, Alabama.

5. Global PM_{2.5}

[28] We tentatively extend our approach to produce a global field of surface $PM_{2.5}$. Figure 6 shows average AOD retrieved from both the MODIS and MISR for January 2001 to October 2002. The largest enhancements are from mineral dust over and downwind of deserts [*Kaufman et al.*, 2005]. Substantial AOD are measured by both MODIS and MISR over eastern China and northern India, associated with industrial pollution and dust storms



Figure 6. Average AOD for January 2001 to October 2002 determined from MODIS and MISR. White space denotes regions with AOD measurements on fewer than either 40% of days for MODIS or 8% of days for MISR.

[*Chu et al.*, 2005; *Kahn et al.*, 2004]. MISR AOD is higher than MODIS AOD over ocean by approximately 0.03 [*Abdou et al.*, 2005]. Over land MODIS AOD is generally higher than MISR AOD by 0.05–0.15, although differences over the Middle East and southwestern United States can reach 0.35. Nonlinearity in the relationship between the infrared and visible surface reflectivity for different land types contributes to the MODIS AOD bias over deserts [*Kaufman et al.*, 2002; *Abdou et al.*, 2005].

[29] Figure 7 shows annual average global remote-sensed PM_{2.5}. Pronounced differences with Figure 6 are driven by large spatial variation in equation (4). The spatial patterns are similar for MODIS and MISR (r = 0.90), with enhancements over major industrial regions and central Africa. MODIS $PM_{2.5}$ tend to be within 2–5 ug/m³ of MISR $PM_{2.5}$, except MODIS exceeds MISR by $10-15 \text{ ug/m}^3$ for parts of China, eastern United States and northern Europe. Both MODIS and MISR show the largest PM2.5 concentrations in Northern India and China, with values in excess of 40 ug/m³. The meteorology, topography and aerosol sources in the Gangetic valley of India favor the development of high PM_{2.5} concentrations [Di Girolamo et al., 2004], contributing to the regional values of 40-50 ug/m³. Ground-based measurements of PM2.5 in China indicate annual average concentrations of 50-100 ug/m³, because of dense population coupled with the use of coal and heavy traffic [Zhang et al., 2004; Oanh et al., 2006]. Low PM2.5 concentrations are found over and downwind of the Sahara, despite high AOD. The weak relation between AOD and surface $PM_{2.5}$ in this region are driven by intense vertical mixing and a large coarse aerosol fraction. Southern Africa displays enhanced surface aerosol concentrations where large seasonal biomass burning occurs [*Formenti et al.*, 2003].

[30] The World Health Organization released in 2005 an air quality guideline for annual mean $PM_{2.5}$ concentrations of 10 ug/m³. We estimate from the remote-sensed $PM_{2.5}$ that at least 70% of the world's population lives in regions that do not meet this standard.

6. Conclusions

[31] We estimated the ground-level concentration of fine particulate mass ($PM_{2.5}$) for January 2001 to October 2002 using space-based measurements from the Moderate Resolution Imaging Spectroradiometer (MODIS) and the Multiangle Imaging Spectroradiometer (MISR) satellite instruments, and additional information from a global chemical transport model (GEOS-CHEM). Remote-sensed $PM_{2.5}$ was compared with surface measurements throughout Canada from the National Air Pollution Surveillance (NAPS) network and throughout the United States from the Air Quality System (AQS).

[32] The spatial variation in annual mean $PM_{2.5}$ exhibited significant agreement with surface measurements when



Figure 7. Average of daily surface $PM_{2.5}$ concentrations for January 2001 to October 2002 determined from MODIS and MISR measurements of AOD. White space denotes regions with AOD measurements on fewer than either 40% of days for MODIS or 8% of days for MISR.

derived from MODIS (r = 0.69, slope = 0.82), and MISR (r = 0.58, slope = 0.57). The daily variation in remote-sensed PM_{2.5} was more consistent with surface measurements in eastern North America (r = 0.5–0.8) than in the western North America (r = 0–0.35) for both MODIS and MISR. We validated the method by deriving PM_{2.5} from AERONET AOD and found a consistent agreement (r = 0.71, slope = 0.69) with the surface measurement of PM_{2.5}. Removing California from the comparison increases the correlation and slope to 0.61 and 0.68 for MISR and 0.80 and 0.89 for AERONET, with little effect on the comparison with MODIS.

[33] We developed an expression to isolate the most important factors affecting the relationship between ground-level $PM_{2.5}$ and AOD. The relative vertical profile of aerosol extinction is the dominant parameter in determining the spatial variation between AOD and $PM_{2.5}$ over North America. Simulation of this information in GEOS-CHEM improves the spatial correlation of remote and surface $PM_{2.5}$ from 0.36–0.37 to 0.58–0.69. In contrast, daily variation in AOD played the major role in accurately representing daily variation in remote-sensed $PM_{2.5}$. Daily variation in parameters such as the relative vertical profile of aerosol extinction or the effective radius was insignificant.

[34] We developed a global map of mean $PM_{2.5}$ using MODIS and MISR measurements of AOD and GEOS-CHEM to relate AOD and $PM_{2.5}$. Large spatial variation in the relationship between AOD and $PM_{2.5}$ contributes to

substantial differences in $PM_{2.5}$ versus AOD. Northern India and East Asia exhibit pronounced $PM_{2.5}$ enhancements of 40–50 ug/m³. Annual mean concentrations of 15– 25 ug/m³ are found over eastern North America, Europe, and the biomass burning region of central Africa. Despite large AOD over and downwind of the Sahara, low $PM_{2.5}$ concentrations (<5 ug/m³) result from strong vertical mixing and a large coarse aerosol fraction.

[35] Satellite measurements of AOD have the potential to provide a unique synopsis of global surface $PM_{2.5}$ concentrations when coupled with additional information from a chemical transport model on the relationship between AOD and $PM_{2.5}$. Development of this capability will depend on the quality of aerosol remote sensing and model simulation of aerosol parameters. Further work should examine the role of internal versus external mixtures in the relationship between AOD and $PM_{2.5}$. Additional constraints on the vertical profile of aerosol extinction from the CALIOP lidar [*Winker et al.*, 2004] should further improve satellite remote sensing of $PM_{2.5}$.

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