Technical Note: Evaluation of the WRF-Chem “Aerosol Chemical to Aerosol Optical Properties” Module using data from the MILAGRO campaign

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Abstract. A comparison between observed aerosol optical properties from the MILAGRO field campaign, which took place in the Mexico City Metropolitan Area (MCMA) during March 2006, and values simulated by the Weather Research and Forecasting (WRF-Chem) model, reveals large differences. To help identify the source of the discrepancies, data from the MILAGRO campaign are used to evaluate the “aerosol chemical to aerosol optical properties” module implemented in the full chemistry version of the WRF-Chem model. The evaluation uses measurements of aerosol size distributions and chemical properties obtained at the MILAGRO T1 site. These observations are fed to the module, which makes predictions of various aerosol optical properties, including the scattering coefficient, $B_{\text{scat}}$; the absorption coefficient, $B_{\text{abs}}$; and the single-scattering albedo, $\sigma_0$; all as a function of time. Values simulated by the module are compared with independent measurements obtained from a photoacoustic spectrometer (PAS) at a wavelength of 870 nm. Because of line losses and other factors, only “fine mode” aerosols with aerodynamic diameters less than 2.5 µm are considered here. Over a 10-day period, the simulations of hour-by-hour variations of $B_{\text{scat}}$ are not satisfactory, but simulations of $B_{\text{abs}}$ and $\sigma_0$ are considerably better. When averaged over the 10-day period, the computed and observed optical properties agree within the uncertainty limits of the measurements and simulations. Specifically, the observed and calculated values are, respectively: (1) $B_{\text{scat}}$, 34.1±5.1 Mm$^{-1}$ versus 30.4±3.4 Mm$^{-1}$; (2) $B_{\text{abs}}$, 9.7±1.0 Mm$^{-1}$ versus 11.7±1.2 Mm$^{-1}$; and (3) $\sigma_0$, 0.78±0.05 and 0.74±0.03. The discrepancies in values of $\sigma_0$ simulated by the full WRF-Chem model thus cannot be attributed to the “aerosol chemistry to optics” module. The discrepancy is more likely due, in part, to poor characterization of emissions near the T1 site, particularly black carbon emissions.

1 Introduction

Radiative aerosol forcing of climate is an area of active study with important implications for climate predictions. These predictions are often provided by global climate models, which in turn rely on parameterizations of the myriad complex physical processes of the climate system. An important step in developing and implementing these parameterizations is the testing of aerosol modules prior to implementation in global climate models. Most of these modules can be categorized as either chemical transport modules (CTM), which calculate the aerosol mass concentrations in space and time, or radiative transfer modules (RTM), which relate aerosol mass to aerosol optical properties and calculate radiative forcing. The evaluation of these modules often uses measurement-model “check points”, as described in Bates et al. (2006); for example, one such check point could be the comparison of aerosol mass measurements with computations of the same obtained from a CTM. If the requisite data are available, the string of calculations that begins with aerosol formation and concludes with radiative forcing can be evaluated at each check point. This approach has the potential to isolate shortcomings in each module or to identify problems with input to the modules, such as emissions.

We adopt a similar approach to evaluate the performance of one aerosol module that computes aerosol optical properties from aerosol chemical properties. The specific module
under scrutiny is part of the chemistry version of the Weather Research and Forecasting model, known as WRF-Chem (Fast et al., 2006; Grell et al., 2005). As noted by Ghan and Schwartz (2007), regional models provide an important test bed for evaluating aerosol process modules, and WRF-Chem represents a well-known example of a regional model being used in this fashion (Fast et al., 2009). Specifically, our evaluation is performed by comparing the module’s simulations of single scattering albedo, $\omega_0$, scattering coefficient, $B_{\text{scat}}$, and absorption coefficient, $B_{\text{abs}}$, to field measurements of these optical properties obtained as part of the MILAGRO field campaign that took place in March 2006 in the Mexico City Metropolitan Area (MCMA). These measurements, as well as the WRF-Chem module, are described below. Evaluation of this module is particularly important because the aerosol optical properties calculated by the module serve as input to other WRF-Chem modules that calculate photolysis rates, such as FAST-J (Barnard et al., 2004; Wild et al., 2000), and shortwave radiative fluxes (Goddard scheme, Chou et al., 1998). Other WRF-Chem modules pertaining to aerosol chemistry have been evaluated by performing idealized box-model studies (e.g. Zaveri et al., 1999, 2008) before their inclusion in WRF-Chem. WRF-Chem has also been evaluated against numerous field campaign measurements, including the International Consortium for Atmospheric Research on Transport and Transformation (ICAR)T/New England Air Quality Study (NEAQS) (Gustafson et al., 2007; Chapman et al., 2009; McKeen et al., 2007), the MILAGRO field campaigns (Fast et al., 2009; Tie et al. 2009), and the Texas Air Quality (TexAQS) 2000 and 2006 field studies (Fast et al., 2006; McKeen et al., 2009; Wilczak et al. 2009). Fast et al. (2009) describe the WRF-Chem configuration used for MILAGRO as well as the emission inventories.

The particular motivation for the evaluation of WRF-Chem’s optical module is illustrated in Fig. 1, which shows a comparison of a full WRF-Chem simulation of the fine-mode $\sigma_0$ (the solid red line) with observations of $\sigma_0$ (the solid blue line) obtained from measurements of $B_{\text{scat}}$ and $B_{\text{abs}}$ made at the T1 site of the MILAGRO campaign (Doran et al., 2007a, b). The T1 site is located at a latitude of 19°43'N, a longitude of 98°08'W, and at an altitude of 2340 m. The wavelength of both the simulations and observations is 870 nm. The measurements in Fig. 1 show a distinct diurnal variation in $\sigma_0$ that is missing in the simulation; additionally, the simulation seriously overestimates the average value of $\sigma_0$ (0.87 and 0.78 for the calculated and observed values, respectively). There are numerous potential sources of these discrepancies, e.g., poor characterization of emissions, problems with the CTM and/or RTM components of WRF-Chem, etc. An evaluation of the “chemistry to optical properties” module can help narrow the range of possibilities.

We note that our evaluation will test the module “as is”, i.e., we do not attempt to improve the module’s performance by adjusting the module’s internal parameters. (Such internal parameters include, for example, the densities assigned to the chemical species.) However, we will briefly evaluate how uncertainties in these parameters affect the final outcome. In this regard our evaluation borrows some features from those found in more formal closure studies (Quinn et al., 1996), where the scrutiny of the uncertainties is vital in determining whether closure, defined as agreement between the calculated and measured properties within experimental uncertainties, is achieved.

2 The WRF-Chem module, input data, and optical data

2.1 WRF-Chem “aerosol chemistry to aerosol optical properties” module

The WRF-Chem “aerosol chemistry to aerosol optical properties” module is based on a sectional approach, because this approach conveniently connects the WRF-Chem chemical module, MOSAIC (Zaveri et al., 2008) to the optical properties module. (From henceforth, unless otherwise mentioned, “module” refers specifically to WRF-Chem’s aerosol “chemical properties to optical properties” module.) The size bins are based on dry physical diameter, $D_p$, with bin widths that increase geometrically. The first bin extends from a lower limit of 0.0390625 µm to an upper limit of 0.078125 µm. These limits increase by powers of two, up to the largest bin, that contains particles that lie in the range, 5 µm < $D_p$ < 10 µm. In each bin, the particles are assumed to be spherical and internally mixed. The use of size bins, spherical particles, and internal mixing are significant simplifications; ideally we would like to compute aerosol optical

![Fig. 1. The solid red line shows the single scattering albedo as calculated by the WRF-Chem model with full prognostic chemistry. The solid blue line represents the observations. (For convenience, we note that day 78 corresponds to 19 March 2006.)](image-url)
properties knowing the chemical composition at every point within an individual aerosol particle of arbitrary size and shape. Single particle approaches are now available for spherical particles (Zaveri et al., 2010; Riemer et al., 2009), but are far too computationally demanding for use in regional scale three-dimensional models, such as WRF-Chem, or in global climate models.

Given the simplifications discussed above, the conversion from chemical to optical properties follows these steps, listed along with simplifying assumptions, as needed:

1. Chemical masses, \( M_{i,j} \), with units \( g/(cm^3 \text{ dry air}) \), and particle number, \( N_i \), with units \#/\( (kg \text{ dry air}) \), are computed for each bin by MOSAIC, where the subscript “\( i \)” denotes bin number (1 through 8) and “\( j \)” the chemical species. Eleven chemical species are considered, which include black carbon (BC), organic mass (OM), water, and various ionic species, such as sulfate and nitrate.

2. For each bin “\( i \)” the masses are converted to volumes, \( V_{i,j} \), with units \( cm^3/(cm^3 \text{ dry air}) \), by dividing by the density of each chemical species, \( \rho_i \), so that \( V_{i,j} = M_{i,j}/\rho_i \). The assumed densities are given in Table 1.

3. The physical diameter assigned to each bin, \( D_{p,i} \), is found by summing over all \( V_{i,j} \) in a bin, assuming spherical particles, so that

\[
D_{p,i} = 2 \left( \frac{1}{11} \sum_{j=1}^{11} V_{i,j}/N_i \right)^{1/3}.
\]

The aerosol size distribution is therefore defined by \( N_i \) and associated \( D_{p,i} \) for each bin.

4. Now that the size distribution is defined, the module calculates the bulk refractive index of the particles in a bin. For this process, we must choose a refractive index mixing rule; these rules have been described in Bond et al. (2006), and references therein. We chose the spherical shell/core configuration, where all species except BC are uniformly distributed within a shell that surrounds a core consisting only of BC. This configuration was selected because as noted in Bond et al., it avoids the artificial absorption enhancement of BC that comes with volume mixing rules, which assume the BC is uniformly distributed throughout the particle. We denote \( m_{s,i} \) and \( m_{c,j} \) as the bulk complex refractive indices of the shell and core, respectively, for the bin “\( i \)”, and let \( m_j \) be the refractive index of each of the chemical constituents “\( j \)”. Then the shell refractive index is given by

\[
m_{s,i} = \frac{\sum_{j=1}^{11} m_j V_{i,j}}{\sum_{j=1}^{11} V_{i,j}}
\]

and the core refractive index is assigned the value of 1.85+0i.71. This value is the midpoint of a range of values thought plausible as presented by Bond and Bergstrom (2006), specified at 550 nm. We note here, however, that Bond and Bergstrom state that this refractive index may be assumed to be constant across the visible spectral region, extending from 400 nm to 700 nm, but it may be much different at ultraviolet and infrared wavelengths. The refractive indices for the shell components are also listed in Table 1.

Shell/core Mie theory (Ackerman and Toon, 1981) is then used to find the absorption efficiency, \( Q_{a,i} \), the scattering efficiency \( Q_{s,i} \), and the asymmetry parameter, \( g_i \), for each bin. The optical properties at 870 nm are found in the usual manner by summing over the size distribution:

\[
\begin{align*}
B_{\text{scat}} &= \sum_{i=1}^{8} N_i Q_{s,i} \pi \left( \frac{D_{p,i}}{2} \right)^2 \\
B_{\text{abs}} &= \sum_{i=1}^{8} N_i Q_{a,i} \pi \left( \frac{D_{p,i}}{2} \right)^2 \\
g &= \frac{\sum_{i=1}^{8} N_i Q_{s,i} \pi \left( \frac{D_{p,i}}{2} \right)^2 g_i}{\sum_{i=1}^{8} N_i Q_{a,i} \pi \left( \frac{D_{p,i}}{2} \right)^2} \\
\sigma_0 &= \frac{B_{\text{scat}}}{B_{\text{scat}} + B_{\text{abs}}}
\end{align*}
\]

where \( g \) is the overall asymmetry parameter.

For the purposes of module testing, the above scheme remains intact, except that measured size distributions and chemical masses are substituted for modeled quantities \( (M_{i,j}, N_i) \) in step one. These measurements are described below.

### 2.2 Aerosol chemical measurements

Aerosol chemical measurements included elemental carbon (EC), aerosol organic carbon (OC) and concomitant OM content, and ionic species. The total mass of aerosols with aerodynamic diameters less than 2.5 \( \mu m \) (called PM\(_{2.5}\) aerosols or PM\(_{2.5}\) mass) was also measured at the T1 site. Elemental carbon and BC are operationally defined (Pöschl, 2003), but for this paper we take EC and BC to be interchangeable. These mass measurements, including the PM\(_{2.5}\) measurement, are used to estimate the fine mode dust content of the aerosol, as will be explained below.
A Sunset Labs OCEC instrument (Birch and Cary, 1996; Doran et al., 2007a, b), using a thermo-optical technique, provided measurements of OC and EC for PM$_{2.5}$ aerosols. The estimate error of these measurements is ±0.2 µg/m$^3$. The organic carbon concentration was converted to OM concentration by multiplying by the factor 1.7 (Aiken et al., 2008), so that OM = 1.7 OC. Inorganic ionic species (e.g., Na$^+$, K$^+$, Ca$^{2+}$, Mg$^{2+}$, Cl$^-$, NO$_3^-$, NO$_2^-$, SO$_4^{2-}$) were measured with a Particle into Liquid Sampler (PILS) instrument (Orsini et al., 2003; Weber et al., 2001), also for PM$_{2.5}$ aerosols. The PILS uses a small amount of water vapor to form water droplets around individual aerosol particles, dissolving water soluble components. The water is collected and analyzed using ion chromatography. This analysis cycle takes about four minutes, thereby producing a semi-continuous time series of aerosol inorganic species. The uncertainty of these measurements is stated as ±10% (Weber et al., 2001).

PM$_{2.5}$ at the T1 site was measured with a tapered element oscillating microbalance (TEOM) instrument, with an estimated uncertainty of ±5%. Figure 2 shows the various mass measurements averaged over the diurnal cycle, in a manner similar to chemical mass measurements presented by Paredes-Miranda et al. (2009) for the MILAGRO T0 site. The procedure averages all the measurements that fall in the time bin delineated by a lower limit of 00:00 LST and an upper limit of 01:00 LST, producing an average value for this hour, and so on for the other 23 h of the day. The display of diurnal averages aids in explaining the variation of aerosol optical properties. These averages are found over a time period extending from day 78 (19 March 2006) through most of day 87 (28 March 2006). Fast et al. (2007) subdivide the meteorology during the MILAGRO campaign into three regimes, and our data span two of these regimes. We segregate the data into one or the other meteorological period; the upper panel shows data for conditions that were mostly clear (19 March 2006 through 23 March 2006, 12:00 LST), while the lower panel is for the regime during which precipitation occurred (23 March 2006, 12:00 LST through 29 March 2006). For convenience, we refer to these two regimes as “clear” and “showery”, respectively. To better show the concentrations of the various species, two graphs are used for each time period. One graph shows PM$_{2.5}$ mass, OM, and “fine mode dust”, while the other graphs show various inorganic species (SO$_4$, NO$_3$, NH$_4$, Cl), crustal materials (Na, Ca, Mg), and EC. The fine mode dust is found by subtracting all known substances (EC, OM, and inorganics) from the PM$_{2.5}$ mass and assuming that this residual is dust. Support for this assumption is found by noting that: (1) the residual is substantially reduced during the showery period, consistent with reduced dust emissions occurring during the wet surface conditions; and (2) a considerable amount of dust was often observed at the T1 site (Querol et al., 2008).

For our analysis we treated the aerosols at the surface as dry. The assumption of a dry aerosol is supported by the work of Moffet et al. (2008a, b), who used single particle mass spectrometry (an aerosol time-of-flight mass spectrometer, ATOFMS) to analyze aerosol chemical and radiative properties, and noted that the radiative microphysical properties displayed no detectable RH dependence, thus indicating dry particles. This is broadly consistent with the measurements of relative humidity (RH) during the MILAGRO campaign, e.g., as reported by Doran et al. (2007a) for the T2 site. They found daytime RH values ranging between 10% and 40% during the clear period, although higher values were found at night and for parts of some days during the showery period.
Fig. 2. Diurnally averaged times series of chemical species and PM$_{2.5}$ mass. The two upper panels show these constituents during the clear period of little precipitation, while the two lower panels show the same constituents during the showery period. For clarity, the mass measurements are broken down into two plots: OM, PM$_{2.5}$, and dust compose one plot, while SO$_4$, NO$_3$, NH$_4$, Cl, crustal materials, and EC compose the second plot. Note that PM$_{2.5}$ mass is significantly lower during the showery period.

We make a few remarks regarding the time variation of the chemical species. First, over the course of the campaign, the mass concentrations are larger during the clear period than the showery period, most likely an effect of precipitation scavenging during the showery period. Second, total mass (PM$_{2.5}$) shows a diurnal trend, with two maxima, one occurring at about 09:00 LST and the other at about 18:00 LST. The first peak results from emissions trapped in the morning stable boundary layer. After about 09:00 LST a convective boundary layer develops, resulting in significant reductions in surface mass concentrations. The secondary peak in PM$_{2.5}$ at 18:00 LST is probably caused by windblown dust. The broad plateau in the inorganic species NO$_3$ and NH$_4$ that occurs between 10:00 LST and 18:00 LST (clear sky period), could be caused by the formation of new secondary inorganic aerosol, as well as residual mass carried over from the previous day. This process was also thought to occur at the T0 site (Paredes-Miranda et al., 2009). Although we cannot identify the cause of these peaks, the spikes in NH$_4$ and NO$_3$ concentrations that appear around noon for the showery period appear to be real, and are seen in the data on three consecutive days, 27 March through 29 March.

2.3 Optical measurements

Optical measurements were made at the MILAGRO T1 site using a photoacoustic spectrometer (PAS), as described in Arnott et al. (1999). This instrument uses sound pressure produced by light absorption in an acoustic resonator to measure aerosol absorption. To find $\sigma_0$, scattering measurements are required, and these also are obtained from the PAS, using reciprocal nephelometry (Rahmah et al., 2006). The use of this particular combination of measurements to find $\sigma_0$ is described in Paredes-Miranda et al. (2009). At the T1 site, the PAS measurements were made at only one wavelength, $\lambda$, of 870 nm. For this study, focusing on $\sigma_0$ at this wavelength is advantageous because we avoid the possibly major complications of dust absorption (Sokolik and Toon, 1999), and organic carbon absorption, which may become significant at wavelengths less than about 600 nm (Bergstrom et al., 2010; Barnard et al., 2008; and references therein). We rely solely on PAS absorption measurements, because these measurements are made without filter substrates. Recent evidence (Lack et al., 2008; Subramanian et al., 2007) suggests that filter-based measurements of absorption, made
in the presence of large amounts of organic carbon, could be highly suspect because of soiling of the filter by the organic component of the aerosol.

Average absorption values were obtained every two minutes and were subsequently averaged over one hour intervals; the 1-h averages are used in this study. Calibration and use of the PAS is described in Lewis et al. (2008), Sheridan et al. (2005), and Arnott et al. (2000). Their experience has led to estimates of the uncertainties in absorption and scattering measurements at the T1 site of 10% and 15%, respectively. These are estimates of systematic error (as opposed to random error) and are not reduced by averaging (Paredes-Miranda et al., 2009). If we assume that these errors work in concert to maximize the error in \( \sigma_0 \) then the uncertainty in the inferred values of \( \sigma_0 \) would be about 6%.

If the particles are large, the magnitude of the scattering measurements will always be less than the true scattering because of difficulties in measuring the forward scattering peak, which becomes significantly more prominent as the particle size increases. However, optical properties of very large particles, with aerodynamic diameters, \( D_a \), greater than about 2 to 3 \( \mu \)m, were not measured by the PAS because of line and inlet losses that reduced the sampling efficiency of large particles to virtually zero. We thus assume that the largest particles that were measured did not exceed 2.5 \( \mu \)m, consistent with PM2.5 measurements. In terms of physical diameter, this cut-off is about 1.9 \( \mu \)m, found using the well-known relationship between physical and aerodynamic diameter for spherical particles, \( D_p = D_a / \sqrt{\rho_{aer}} \) (e.g., Shaw et al., 2007), where \( \rho_{aer} \) is the density of the aerosol, taken to be 1.8 g/cm\(^3\). Because coarse mode particles could not be measured, our \( \sigma_0 \) values should be considered as “fine mode” values.

### 2.4 Size distribution

Aerosol size distributions were measured at the surface using a scanning mobility particle sizer (SMPS; this instrument is a TSI Model 3936 teamed with a TSI Model 3081 differential mobility analyzer and a 3025A condensation particle counter; http://www.tsi.com). As configured during the MI-LAGRO campaign, this instrument sized particles between 0.014 and 0.74 \( \mu \)m in electric mobility diameter. As noted by DeCarlo et al. (2004), the electric mobility diameter is equivalent to the physical diameter for spherical particles. Because the upper limit of the size distribution measurements is too small to capture the larger particles sensed by the PAS and chemical measurement equipment, it is necessary to extrapolate the SMPS measurement to larger diameters.

Figure 3 illustrates this process. The upper panel shows a typical size distribution (the blue dots) measured during the clear period, while the lower panel shows a similar size distribution taken during the showery period. The two distributions differ in shape for particles larger than about 0.5 \( \mu \)m, at which point the size distribution increases for larger par-

![Fig. 3](http://www.tsi.com)  
**Fig. 3.** The blue dots represent the measured size distribution, as measured by the SMPS. The size distribution is an hourly average. The red curves show fits to these data.

Particles during the clear period and decreases for the showery period. These characteristic shapes were fairly consistent for all the hourly averaged size distributions in each respective period. The presence of larger particles for the clear period could be due to windblown dust, which was presumably being suppressed by rainfall during the showery period.

For the showery times, we found that a log-normal function fit the size distribution quite well, and we used this function to extrapolate the distribution to sizes larger than 0.74 \( \mu \)m. The fit distribution is shown by the red line. For the clear periods, we used two log-normal distributions, once of which was fit to particles less than 0.5 \( \mu \)m in diameter, and the other to the larger sizes. This combined distribution is also shown in Fig. 3. For both the clear and showery periods, both distributions tend to zero as the size increases, thereby simulating, perhaps crudely, the sampling inefficiencies of the chemical and optical instruments. The credibility of these extrapolated size distributions is bolstered by comparing total aerosol volume derived from them with the volume estimates, \( V_{i,j} \), obtained from the aerosol mass measurements (e.g., see Sect. 2.1). The \( V_{i,j} \) are summed over all bins “i” and all chemical constituents “j” to yield total aerosol volume, \( V_m \), where the subscript “m” denotes that the volume has been derived from the mass measurements. For this procedure we use hourly averaged values. By integration of the hourly averaged size SMPS distributions for all \( D_p < 2.0 \mu \)m, we obtain another measure of the aerosol volume denoted \( V_s \). Figure 4 shows a comparison of the two volumes, segregated by time into “clear” (black dots) and “showery” (red dots) periods. Given the approximations involved, the correlation is satisfactory. We note in passing that during the showery period, when aerosol mass is generally reduced, both volumes are generally smaller than during the clear period, as would be expected.
observed values are shown by the blue line. We note that there are several time periods with missing size distribution and/or chemical mass data, and for these periods no calculations are possible. One long stretch occurs from the end of day 84 (25 March 2006) though most of day 85.

Focusing first on $B_{\text{abs}}$, the calculated and observed $B_{\text{abs}}$ values exhibit similar diurnal patterns, with a large peak occurring between 06:00 and 08:00 LST and much smaller values at other times. These diurnal maxima correlate well with the diurnal maxima in BC concentration seen in Fig. 2, suggesting that fluctuating BC concentrations control most of the absorption at 870 nm. Small absorption contributions by dust and OM are possible, but the absorption signal of these components is probably too small to detect at 870 nm. Overall, the module exhibits reasonable skill in predicting $B_{\text{abs}}$, but has a tendency to overestimate the observed values. The regression line between simulations and observations is $B_{\text{abs, calculated}} = -1.3 \text{ Mm}^{-1} + 1.34 \cdot B_{\text{abs, observed}}$ with a correlation coefficient ($r^2$) of 0.82. When averaged over the entire comparison period, the calculated and observed values of $B_{\text{abs}}$ are 11.7 $\text{Mm}^{-1}$ and 9.7 $\text{Mm}^{-1}$, respectively.

For $B_{\text{scat}}$, we see that the agreement between calculated and observed values is more problematic; the regression line is $B_{\text{scat, calculated}} = 14.8 \text{ Mm}^{-1} + 0.46 \cdot B_{\text{scat, observed}}$ with an $r^2 = 0.16$. The module does calculate larger $B_{\text{scat}}$ values during the clear period and smaller values during the showery time, consistent with the optical data and with the decreased aerosol volume during the showery period (Fig. 3), but there is a tendency for the calculated peaks to occur a few hours before the observed ones, especially during the clear period. Interestingly, when averaged over the entire comparison periods, the calculated and observed values are remarkably similar, 30.4 $\text{Mm}^{-1}$ and 34.1 $\text{Mm}^{-1}$, respectively. We will discuss the uncertainty of these values below. The averaged extinction coefficient, $B_{\text{ext}}$, is the sum of the averages of $B_{\text{scat}}$ and $B_{\text{abs}}$, or 42.1 $\text{Mm}^{-1}$ and 43.8 $\text{Mm}^{-1}$, for calculated and observed values, respectively.

Figure 6 shows time series of calculated and observed $\sigma_0$ values (Eq. 2) at 870 nm. A distinct diurnal pattern is evident in the observed time series for $\sigma_0$. During the course of a day, $\sigma_0$ has a pronounced minimum at about 06:00 LST and a broad maximum around 15:00 LST. We see that despite the difficulty the module has in predicting the daily pattern of $B_{\text{scat}}$, the diurnal behavior of the observations is approximately captured by the calculations of $\sigma_0$. The correlation ($r^2$) between observed and calculated values is 0.56; with an associated regression line of $\sigma_0,\text{ calculated} = 0.13 + 0.79 \cdot \sigma_0,\text{ observed}$. The mean values of $\sigma_0$ over the course of the comparison period are 0.74 and 0.78, for the calculated and observed values, respectively.

For the MILAGRO T1 site, the large daily swings in $B_{\text{abs}}$ govern the diurnal behavior of the $\sigma_0$, in part because $\sigma_0$ values are more sensitive to changes in $B_{\text{abs}}$ than $B_{\text{scat}}$. (By considering $\partial \sigma_0/\partial B_{\text{scat}}$ and $\partial \sigma_0/\partial B_{\text{abs}}$, and using typical values for $B_{\text{abs}}$, $B_{\text{scat}}$, and $\sigma_0$, we find that $\sigma_0$ is three times
Fig. 5. The top panel shows the absorption coefficient, $B_{abs}$, while the bottom panel shows the scattering coefficient, $B_{scat}$. The blue lines indicate the observations, while the red dots show simulations of these coefficients derived from the WRF-Chem “chemical to optical properties” module. Hourly averages are shown. The dashed red line that connects the dots aids in the comparison between the simulations and observations. Note that there are significant time periods when missing data prevented a simulation from taking place; for example, the time span from the end of day 84 continuing on through most of day 85. We again note for convenience that day 78 is 19 March 2006. The vertical, bold dashed line that occurs at time 82.5 separates the clear (julian day < 82.5) and showery (julian day ≥ 82.5) periods.

The module performance depicted in Fig. 6 shows a distinct improvement over the full WRF-Chem simulation in Fig. 1. This implies that most of the discrepancy in that figure cannot be attributed to the module evaluated here.

Over the entire comparison period the averaged $\sigma_0$ for the calculations and observations are 0.74 and 0.78, respectively. Marley et al. (2009) report a mean observed value of $\sigma_0$ at 550 nm of 0.68 at the T1 site and a diurnal pattern similar to that reported here (Figure 1), over a time period extending from 1 March 2006 through 29 March 2006 (e.g., see Figure 3 in Marley et al.). This value is lower than that measured at 870 nm, suggesting enhanced absorption at the lower wavelengths, perhaps attributable to dust (Bergstrom et al., 2010; Bergstrom et al., 2007) or organic carbon (Bergstrom et al., 2010; Barnard et al., 2008; Kirchstetter et al. 2004).

For the sake of comparison, it is interesting to show the aerosol optical properties for the full, prognostic WRF-Chem run. These are shown in the fourth column of Table 2. The single scattering albedo, for the full 10-day period, is 0.87, as noted above. The prognostic WRF-Chem simulation substantially underpredicts $B_{abs}$ for all time periods, and overpredicts $B_{scat}$ for the full time period as well as for the showery period. For the clear period, both observed and
the simulated and observed PM. It is too small. For example, during the showery period, PM 2.5 is significantly too large, and vice versa for the predicted PM 2.5. There are significant differences (0.05 and 0.07) between the observed and calculated PM 2.5 concentrations, both measured and as predicted by WRF-Chem. This table also shows comparisons for the clear and showery periods. Optical properties that do not agree within estimated uncertainties are shown in boldface. The wavelength associated with all of properties listed here is 870 nm.

Table 2. Mean values of observed and calculated aerosol optical properties for the period 19 March through 28 March 2006. The calculated properties are shown for the module only, as well as the full, prognostic WRF-Chem model. This table also shows comparisons for the clear and showery periods. Optical properties that do not agree within estimated uncertainties are shown in boldface. The wavelength associated with all of properties listed here is 870 nm.

<table>
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<th>Optical property</th>
<th>Observations (with uncertainties)</th>
<th>WRF-Chem module (with uncertainties)</th>
<th>Full prognostic WRF-Chem model</th>
<th>Full prognostic WRF-Chem with observed BC</th>
<th>Time period</th>
</tr>
</thead>
<tbody>
<tr>
<td>σ0</td>
<td>0.78±0.05</td>
<td>0.74±0.03</td>
<td>0.87</td>
<td>0.78</td>
<td>Full</td>
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<tr>
<td>Bscat</td>
<td>34.1±5.1 Mm⁻¹</td>
<td>30.4±3.4 Mm⁻¹</td>
<td>46.4 Mm⁻¹</td>
<td>46.5 Mm⁻¹</td>
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<tr>
<td>Bab</td>
<td>9.7±1.0 Mm⁻¹</td>
<td>11.7±1.2 Mm⁻¹</td>
<td>5.6 Mm⁻¹</td>
<td>11.1 Mm⁻¹</td>
<td>&quot;</td>
</tr>
<tr>
<td>σ0</td>
<td>0.77±0.05</td>
<td>0.74±0.03</td>
<td>0.85</td>
<td>0.72</td>
<td>Clear</td>
</tr>
<tr>
<td>Bscat</td>
<td>38.7±5.8 Mm⁻¹</td>
<td>38.1±4.1 Mm⁻¹</td>
<td>37.8 Mm⁻¹</td>
<td>38.0 Mm⁻¹</td>
<td>&quot;</td>
</tr>
<tr>
<td>Bab</td>
<td>11.2±1.1 Mm⁻¹</td>
<td>14.6±1.5 Mm⁻¹</td>
<td>5.6 Mm⁻¹</td>
<td>13.0 Mm⁻¹</td>
<td>&quot;</td>
</tr>
<tr>
<td>σ0</td>
<td>0.79±0.05</td>
<td>0.74±0.03</td>
<td>0.90</td>
<td>0.86</td>
<td>Showery</td>
</tr>
<tr>
<td>Bscat</td>
<td>28.7±4.3 Mm⁻¹</td>
<td>21.2±3.5 Mm⁻¹</td>
<td>56.7 Mm⁻¹</td>
<td>56.4 Mm⁻¹</td>
<td>&quot;</td>
</tr>
<tr>
<td>Bab</td>
<td>8.0±0.8 Mm⁻¹</td>
<td>8.3±1.3 Mm⁻¹</td>
<td>5.5 Mm⁻¹</td>
<td>8.9 Mm⁻¹</td>
<td>&quot;</td>
</tr>
</tbody>
</table>

Table 3. Averaged concentration of PM2.5 and BC for the periods indicated. The label “WRF-Chem” means that these are the concentrations predicted by the full, prognostic WRF-Chem model.

<table>
<thead>
<tr>
<th>Time period</th>
<th>PM2.5 (WRF-Chem) µg/m³</th>
<th>PM2.5 (observed) µg/m³</th>
<th>BC (WRF-Chem) µg/m³</th>
<th>BC (observed) µg/m³</th>
</tr>
</thead>
<tbody>
<tr>
<td>all</td>
<td>32.7</td>
<td>28.9</td>
<td>0.70</td>
<td>1.54</td>
</tr>
<tr>
<td>clear</td>
<td>25.8</td>
<td>38.3</td>
<td>0.71</td>
<td>1.98</td>
</tr>
<tr>
<td>showery</td>
<td>40.0</td>
<td>21.2</td>
<td>0.68</td>
<td>1.19</td>
</tr>
</tbody>
</table>

calculated Bscat values are about the same. We then ask, why is Bab so grossly underpredicted? Table 3 shows PM2.5 and BC concentrations, both measured and as predicted by WRF-Chem. A comparison of the BC concentrations reveals that the amount of BC found in the WRF-Chem simulation is far lower than the measurements; for example, for the full time period, the BC concentration is 0.70 µg/m³ for WRF-Chem, yet the measured value is 1.54 µg/m³. Because BC is a primary emmission that is not altered significantly in the atmosphere, we attribute WRF-Chem’s poor simulation of BC to the emissions inventory that does not contain enough BC.

This table also shows that on an overall basis, the prediction of PM2.5 is similar for WRF-Chem (32.7 µg/m³) and the observations (28.9 µg/m³), but major differences occur in the clear and showery period. During the clear period, WRF-Chem significantly underpredicts the PM2.5 mass, and the opposite is true in the showery period. We cannot yet explain this behavior. Because PM2.5 is closely related to the scattering (at 870 nm), when the predicted PM2.5 is too large relative to the observations, the predicted Bscat is similarly too large, and vice versa for the predicted PM2.5, when it is too small. For example, during the showery period, the simulated and observed PM2.5 values are 40.0 µg/m³ and 21.2 µg/m³, respectively, while the simulated and observed Bscat is 56.7 Mm⁻¹ and 28.7 Mm⁻¹, respectively. A doubling of PM2.5 leads to a doubling in the scattering. For the showery period, the discrepancy between predicted and observed PM2.5 significantly influences the scattering, and therefore the value of σ0. If we calculate σ0 using the observed PM2.5 (less scattering) in place of the WRF-Chem PM2.5 (more scattering), we find that σ0 drops by about 0.09.

To bolster the conjecture that the specified emissions of BC are too low, we start with the chemical concentrations as simulated by WRF-Chem. We make a single change to these concentrations: we replace the simulated BC concentration by the observed concentration of BC. When this new input is fed to the module, the overall σ0 value is now 0.78, the same value as the observations. However, during the clear and showery periods, there remain sigificant differences (0.05 and 0.07) between the observed and calculated σ0 values. The various optical properties as simulated by the module, using WRF-Chem predicted chemical concentrations with the predicted BC replaced by the measured BC, are shown in the fifth column of Table 2.
Aerosol shape and morphology. The module treats aerosol radiative forcing, and the actual aerosols found in MCMA, as shown by scanning and transmission electron micrographs, are much more complex (Doran et al., 2008; Adachi and Buseck, 2008; Adachi et al., 2007). A detailed treatment of aerosol shells with non-spherical, and possibly tortured shapes, random inclusions of BC, and complex morphologies is not possible in current models, and simplification is thus required. Fuller et al. (1999) attempt to account for the random positioning of BC encapsulated by a spherical sulfate shell, and the resulting BC specific absorption is less than pure shell/core configuration by about 15%. With that study in mind we estimate a random error of ±15% to $B_{\text{scat}}$ and $B_{\text{abs}}$ to account for the departure from a shell/core configuration.

3.2 Uncertainties

Measurement and modeling uncertainties need to be assessed to obtain credible model evaluations (e.g., Bates et al., 2006). There are a number of potential sources of random error, including the assumptions made in the module, errors in the input data, and sampling errors associated with the measurements, as well as possible systematic errors that may be difficult to specify. In the following section we give quantitative estimates of likely random errors when that is possible, and follow with a brief discussion of two potential sources of systematic errors.

We subdivide the random errors into two classes: those associated with module assumptions, and those associated with measurements. We first discuss the assumptions made in the module. These include:

1. Aerosol shape and morphology. The module treats aerosols as spherical particles with a shell/core configuration, but the actual aerosols found in MCMA, as shown by scanning and transmission electron micrographs, are much more complex (Doran et al., 2008; Adachi and Buseck, 2008; Adachi et al., 2007). A detailed treatment of aerosol shells with non-spherical, and possibly tortured shapes, random inclusions of BC, and complex morphologies is not possible in current models, and simplification is thus required. Fuller et al. (1999) attempt to account for the random positioning of BC encapsulated by a spherical sulfate shell, and the resulting BC specific absorption is less than pure shell/core configuration by about 15%. With that study in mind we estimate a random error of ±15% to $B_{\text{scat}}$ and $B_{\text{abs}}$ to account for the departure from a shell/core configuration.

2. Assumptions regarding chemical species density. In the module each chemical constituent is assigned a density. For BC, Bond and Bergstrom (2006) state the plausible density range extends from 1.7 to 1.9 g/cm$^3$, and we have chosen the midpoint of this range as our density value for BC. Using their range to define the extent of the possible error, we estimate the possible error to be about ±5%. Because little information is available about the density ranges of other substances, we assume that this error is applicable to other densities as well, and we consider the error random. Note, however, that if there is discrepancy between the assigned density and the actual, average density, this would result in a systematic error of unknown magnitude.

3. Assumptions regarding the refractive index. For BC, Bond and Bergstrom (2006) defined a range of plausible complex refractive indexes at 550 nm, with the real part, $n$, ranging from 1.75 to 1.95, and the complex part, $k$, ranging from 0.63 to 0.79. Again, we chose the midpoint of these values and, consistent with their range of values, assumed random errors of ±5% and ±11% for $n$ and $k$, respectively. For various organic compounds, Kanakidou et al. (2005) report ranges of $n$ extending from about 1.35 to 1.75 (for a wavelength range of about 300 nm to 800 nm), but with most compounds falling in a smaller interval, from 1.40 and 1.55. Accordingly, we take the uncertainty for $n$ of OM to be ±5% and assign this same level of uncertainty to $n$ for inorganic compounds. These latter compounds are assumed to be non-absorbing and $k$ is set equal to zero. For the refractive index of dust, Mishra and Tripathi (2008) show a range of possible values at 870 nm, with a large percent

<table>
<thead>
<tr>
<th>Parameters/forcing</th>
<th>base case (measured $\sigma_0$ and $B_{\text{ext}}$)</th>
<th>WRF-Chem module calculated $\sigma_0$</th>
<th>$\tau$ scaled by $(B_{\text{ext,calculated}})/ (B_{\text{ext,observed}})$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\sigma_0$ (870 nm)</td>
<td>0.78</td>
<td>0.74</td>
<td>0.78</td>
</tr>
<tr>
<td>$g$ (870 nm)</td>
<td>0.58</td>
<td>0.58</td>
<td>0.58</td>
</tr>
<tr>
<td>$\tau$ (870 nm)</td>
<td>0.12</td>
<td>0.12</td>
<td>0.115</td>
</tr>
<tr>
<td>$\sigma_0$ (500 nm)</td>
<td>0.814</td>
<td>0.78</td>
<td>0.814</td>
</tr>
<tr>
<td>$g$ (500 nm)</td>
<td>0.60</td>
<td>0.60</td>
<td>0.60</td>
</tr>
<tr>
<td>$\tau$ (500 nm)</td>
<td>0.247</td>
<td>0.247</td>
<td>0.237</td>
</tr>
<tr>
<td>EAE</td>
<td>1.3</td>
<td>1.3</td>
<td>1.3</td>
</tr>
<tr>
<td>AAE</td>
<td>1.0</td>
<td>1.0</td>
<td>1.0</td>
</tr>
<tr>
<td>TOA forcing (W/m$^2$)</td>
<td>−2.3</td>
<td>−0.86</td>
<td>−2.2</td>
</tr>
</tbody>
</table>

Table 4. SBDART inputs for three cases: (1) a “base case” that uses measured $\sigma_0$ and $B_{\text{ext}}$ to calculate top of atmosphere (TOA) aerosol forcing, (2) a case that uses the “WRF-Chem optical module” calculated $\sigma_0$ in place of the measured $\sigma_0$, and (3) a case that uses the measured $\tau$ scaled by the ratio $B_{\text{ext,calculated}}/ B_{\text{ext,observed}}$. Also shown are the TOA forcings, averaged over 24 h at the equinox. These TOA forcings are listed in the last row. The spectral surface albedo is from Coddington et al. (2008). The parameters that are changed from the base case are in boldface type.
variation in the complex part and much less variation in the real part. Prasad and Singh (2007), and references therein, discuss possible values for the refractive index of dust suspended over the Indo-Gangetic plain. Using AERONET data, the range in refractive indices for dusty days was, for \( n \), 1.51 to 1.60, and for \( k \), 0.0011 to 0.0033, at a wavelength of 873 nm. Lacking more precise information about optical properties of dust in the MCMA area, we take the refractive index of dust to be the midpoint of these ranges, with errors of \( \pm 5\% \) and \( \pm 100\% \) for the real and imaginary parts, respectively. The large percent error on the imaginary component is not a significant problem in the current evaluation because the absorption is dominated by black carbon.

4. The conversion of organic carbon mass to organic matter mass. Various studies (Aiken et al., 2008; DeCarlo et al., 2008; Malm et al., 2005; Turpin and Lim, 2001) have suggested values of the conversion factor ranging from 1.4 to 2.3, depending on the type of aerosol considered. The most relevant study for our work is probably that of Aiken et al. (2008), who reported an average value of 1.71 at the T0 site. Given that T1 is also primarily urban in character, we assumed a value of 1.7, with an uncertainty of \( \pm 0.2 \).

Measurement uncertainties come into play with both input data (e.g., chemical masses) and the PAS. Some of these uncertainties have been briefly mentioned above but for convenience we repeat them here. These errors include:

1. Errors in the PAS measurements. The magnitudes of these errors, for the PAS instrument at the T1 site, are \( \pm 15\% \) and \( \pm 10\% \) for \( B_{\text{scat}} \) and \( B_{\text{abs}} \), respectively.

2. Sampling efficiency of the PAS. It is assumed that particles with aerodynamic diameter larger than 2 to 3 \( \mu m \) were not sampled. However, the sampling efficiency was not precisely quantified, thus generating an error of unknown but presumably small magnitude.

3. Errors in the measurements of \( \text{PM}_{2.5} \) chemical masses used as input data. For the PILS instrument used to sample inorganic species, the uncertainty of the measurements is given as \( \pm 10\% \) (Weber et al., 2001). The estimated uncertainty of the OC/EC instrument is \( \pm 0.2 \mu g/m^3 \), and the uncertainty in the \( \text{PM}_{2.5} \) mass measurements from the TEOM instrument is \( \pm 5\% \).

4. Size distribution measurement errors. Errors in number concentrations are \( \pm 10\% \) for each size channel, similar to what has been reported for other size measuring instruments (e.g., Wang et al., 2002). As previously mentioned, there is an additional, unknown error because of the extrapolation necessary to extend the size distribution from 0.735 \( \mu m \) to larger sizes. The magnitude of this error can be estimated by changing the magnitude of the portion of the size distribution that is extrapolated. For example, if the extrapolated part of the volume distribution in the top (bottom) panel of Fig. 3 is multiplied by a factor of two, the scattering increases by 24\% (5\%).

If these errors are indeed random, then when averages are taken some error cancellation will occur. We estimated the overall random error by taking a Monte Carlo approach (Bettington and Robinson, 1992). We assumed all the errors were uncorrelated, and perturbed the variables in question (i.e., density, refractive index, morphology error, etc.) using normal deviates, where the standard deviation of the deviate distribution is assumed to be the random error estimates given above. We then ran the module 50000 times and found the distributions of \( B_{\text{scat}} \), \( B_{\text{abs}} \), and \( \sigma_0 \). The widths of the distributions, the standard deviations, were taken as estimates of the random errors in the results. This technique is very rapid, avoids a tedious propagation of error analysis, and has been used to evaluate errors in inferred OM mass absorption coefficients (Barnard et al., 2008). The resulting random errors are \( \pm 3.4 \) \( \text{Mm}^{-1} \) for \( B_{\text{scat}} \) and \( \pm 1.2 \) \( \text{Mm}^{-1} \) for \( B_{\text{abs}} \). For \( \sigma_0 \) the random error is \( \pm 0.03 \).

Systematic errors are more difficult to deal with because they can be difficult to specify. Averaging in general does not reduce systematic errors, although some error reduction can occur because systematic errors of different sign will at least partially cancel. However, systematic errors might explain certain discrepancies between measured and observed optical properties. For example, within each of the 8 model bins, the aerosols are assumed to be internally mixed. Based on a detailed examination of individual particles sampled in the Mexico City plume, Adachi and Buseck (2008) assert that the internal mixing assumption is “relatively reliable for modeling,” but Doran et al. (2008) find that at the T1 site, coating of BC particles progresses rapidly during the daylight hours but is limited or even absent during the night. Uncoated BC has a specific absorption of about 7.5 \( \text{m}^2/\text{g} \) at 550 nm (Bond and Bergstrom, 2006), and at the T1 site, the measured specific absorption is close to this value in the early morning hours and increases to about 11 \( \text{m}^2/\text{g} \) by noon (Doran et al., 2008). We use this finding to estimate the error in neglecting externally mixed BC; the estimated error is 11 \( \text{m}^2/\text{g} \) \( \div \) 7.5 \( \text{m}^2/\text{g} \) \( \approx \) 1.5, or up to 50\% error in the calculation of \( B_{\text{abs}} \), if all the BC is externally mixed. This error is systematic, and would occur in the morning where a significant fraction of the BC load is probably externally mixed. In fact, examination of the bottom panel of Figure 5 shows that there are many cases of early morning spikes in \( B_{\text{abs}} \) where the calculated \( B_{\text{abs}} \) (assumed internally mixed) exceeds the measured \( B_{\text{abs}} \) (perhaps external mixed) by a large amount. A striking case is the absorption spike on day 80 (21 March 2006), when the observed and calculated values of \( B_{\text{abs}} \) are about 35 \( \text{Mm}^{-1} \) and 55 \( \text{Mm}^{-1} \), respectively.
Another possible source of systematic error is the attribution of the residual between PM$_{2.5}$ mass and the sum of all other aerosol mass to dust. We examine the possible magnitude of this error by modifying the density and refractive index of the material we assume is dust. Instead of assigning this material the properties of dust (density = 2.6 g/cm$^3$; refractive index = 1.55 + i0.002), we let the density be 1.8 g/cm$^3$ and set the complex part of the refractive index equal to zero. This calculation reveals that the average value of $B_{\text{scat}}$ increases from 30.4 Mm$^{-1}$ to 31.1 Mm$^{-1}$, while $B_{\text{abs}}$ decreases from 11.7 Mm$^{-1}$ to 11.0 Mm$^{-1}$. The single scattering albedo changes from 0.74 to 0.76. These are not large errors. The fact that the aerosol volumes derived from the chemical data and SMPS are similar (Fig. 4) further suggests that possible errors arising from the effects of fine mode dust are not likely to be significant.

The measurement uncertainties for $B_{\text{abs}}$ and $B_{\text{scat}}$ are 10% and 15%, respectively, which leads to an uncertainty of about 6% in $\sigma_0$. The second and third columns of Table 2 show the calculated and observed value of these quantities, along with associated uncertainties. Averaged over the full ten-day period, the simulations and measurements agree within estimated uncertainties. The agreement is not as good if averages are taken over the two distinct meteorological regimes that subdivide the full ten-day period, but the agreement between observed and calculated values still falls within the stated uncertainty ranges for $B_{\text{scat}}$ for both periods and for $B_{\text{abs}}$ during the showery period. These averages are also shown in Table 2.

The differences between calculated and observed aerosol optical properties as listed in Table 2 will lead to errors in aerosol direct radiative forcing. These errors can be estimated using the method described in McComiskey et al. (2008). We define the top of atmosphere (TOA) aerosol broadband forcing, $F$, in the conventional manner, $F = (f_a ↓ - f_a ↑) - (f_0 ↓ - f_0 ↑)$, where $(f_a ↓ - f_a ↑)$ denotes the net instantaneous downwelling shortwave broadband flux at the TOA in the presence of aerosols, and $(f_0 ↓ - f_0 ↑)$ is the net instantaneous downwelling TOA flux without aerosols. Following McComiskey et al., we find the average solar forcing, $F_{\text{S,ave}}$, where the average is taken over 24 h at the equinox. The flux calculations are made using the SBDART model (Ricchiazzi et al., 1998), with atmospheric conditions typical for the T1 site. The base case aerosol optical properties represent plausible values for the T1 site, and are specified at 870 nm: optical depth, $\tau = 0.12$; asymmetry parameter, $g = 0.58$; extinction angstrom exponent (EAE) = 1.3; absorption angstrom exponent (AAE) = 1.0, and $\sigma_0 = 0.78$. The value specified for $\sigma_0$ is the observed surface value, and we assume that it is constant throughout the depth of the atmosphere. The surface spectral albedo for the T1 site is from the analysis of Coddington et al. (2008).

SBDART input variables, and the TOA forcings are listed in Table 4, for: (1) case one, the base case, (2) case two, same as the base case, except $\sigma_0$ is set equal to the WRF-Chem module calculated value of 0.74 instead of the observed value of 0.78, and (3) case three, same as the base case, except the observed $\tau$ = 0.12 is scaled by the factor $B_{\text{ext,calculated}}/B_{\text{ext,observed}}$ = 42.1 Mm$^{-1}$/43.8 Mm$^{-1}$ = 0.96, where $B_{\text{ext,calculated}}$ and $B_{\text{ext,observed}}$ are the average extinction coefficients calculated from the module and the observations, respectively. Here we assume that the surface scaling of extinction can be uniformly extrapolated throughout the atmosphere.

For the base case, the TOA forcing is $-2.3$ W/m$^2$. For case two, the forcing is $-0.86$ W/m$^2$, a difference of about 1.4 W/m$^2$ from the base case. The difference between the base case and case three is negligible. McComiskey et al. (2008) state that the largest contributor to forcing uncertainty is $\sigma_0$ and this is consistent with our results. The 1.4 W/m$^2$ difference is somewhat greater than the maximum uncertainty in TOA forcing, 1.1 W/m$^2$, as estimated by McComiskey et al. (2008) (e.g., see Table 4 in McComiskey et al.). However, when making this comparison, we must be mindful that they considered other sources of uncertainty, in addition to just $\sigma_0$ and $\tau$.

### 4 Conclusions

This study was originally motivated by the failure of the WRF-Chem model, run with full prognostic chemistry, to simulate $\sigma_0$ satisfactorily over the MILAGRO T1 site over a 10-day period. To help identify the source of the discrepancy, we extracted the WRF-Chem “aerosol chemistry to aerosol optical properties” module from the full WRF-Chem code and used observed (rather than simulated) values of aerosol chemical species and size distributions as input to the module. We then tested its ability to simulate the observed T1 aerosol optical properties $B_{\text{scat}}$, $B_{\text{abs}}$, and $\sigma_0$ at a wavelength of 870 nm.

Although some difficulties with the “aerosol chemistry to aerosol optical properties” module were found, we conclude that any shortcomings in the module are unlikely to have been a major factor in the discrepancies with the observed values of $\sigma_0$ found using the full WRF-Chem model. A more significant source of error likely is the difficulty of specifying emissions accurately. For example, the full WRF-Chem model gave an average value of $\sigma_0$ of 0.87. Fast et al. (2009) showed that simulated BC was usually lower than observed, particularly between 11:00 and 16:00 UTC, with a bias of $-1.3$ µg/m$^3$ during the entire MILAGRO campaign. When the observed BC mass concentrations at T1 were substituted into the full simulation (not shown), replacing the simulated values derived from the emissions inventory estimates used as model input, the simulated mean value for $\sigma_0$ decreases to 0.78, which is a the same as the observed mean value of 0.78. Moreover, a significant portion of the diurnal variation in $\sigma_0$ was then simulated. This suggests that poor specification of BC emissions may be the primary cause for
the poor $\sigma_0$ simulation by WRF-Chem. However, for the showery case, WRF-Chem overpredicts PM$_{2.5}$ mass, leading to an overprediction of scattering and therefore $\sigma_0$. This shows that predicted PM$_{2.5}$ mass also plays a role in determining aerosol optical properties. When the aerosol module alone was evaluated using measured chemical species and size distributions as input, the average simulated value of $\sigma_0$ was 0.74, and the diurnal variation was captured reasonably well.

On an hour-by-hour basis the aerosol module does not perform satisfactorily in predicting $B_{\text{scat}}$, ($r^2 = 0.16$), but does considerably better for $B_{\text{abs}}$, ($r^2 = 0.82$) and $\sigma_0$ ($r^2 = 0.56$). The observed (and pronounced) diurnal patterns in $B_{\text{abs}}$ and $\sigma_0$ were approximately captured by the module (see Figs. 5 and 6), although the module tends to predict higher values of $B_{\text{abs}}$ in the morning (around 06:00 LST), when the concentration of BC is largest. We suggest this may arise because the module’s assumption of full internal mixing in each bin is not appropriate during the morning hours.

The module shows better skill in simulating all three optical properties when averages over the 10-day period are considered, as summarized in Table 2. This table shows that: (1) for $B_{\text{scat}}$, the observed and calculated values are $34.1 \pm 5.1 \, \text{Mm}^{-1}$ and $30.4 \pm 3.4 \, \text{Mm}^{-1}$, respectively; (2) for $B_{\text{abs}}$, the observed and calculated values are $9.7 \pm 1.0 \, \text{Mm}^{-1}$ and $11.7 \pm 1.2 \, \text{Mm}^{-1}$; and (3) the observed and calculated $\sigma_0$ values are $0.78 \pm 0.05$ and $0.74 \pm 0.03$. These values include estimated uncertainties in the averages due to random error, as discussed in Sect. 3.2. Our estimate of the uncertainty in averaged, TOA aerosol forcing, attributable to module inaccuracies in calculating $\sigma_0$ and $B_{\text{ext}}$, is 1.4 W/m$^2$. The bulk of this uncertainty is induced by the difference between calculated and observed $\sigma_0$.

For climate simulations where hour-by-hour variations are less important, the current “aerosol chemistry to aerosol optical properties” module may be satisfactory and could prove to be quite useful. Significant improvement to the module simulations may be realized when a two-dimensional size distribution is used that considers the size of the aerosol as a whole, as well as the size of the BC inclusions (Zaveri et al., 2010; Oshima et al., 2009). For studies of more episodic events, additional work will be required to identify and correct the current shortcomings of the module, particularly in regard to the scattering calculations. Further testing, using data from other locations, would also be useful to determine how well the module performs with other mixtures of aerosols and higher relative humidity. Because the aerosols in the MCMA are dry, we do not know how well the module would perform in areas with large relative humidity and concomitant hydroscopic growth.

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