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Characterization of atmospheric aerosol in the US Southeast from ground- and space-based measurements over the past decade

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Abstract

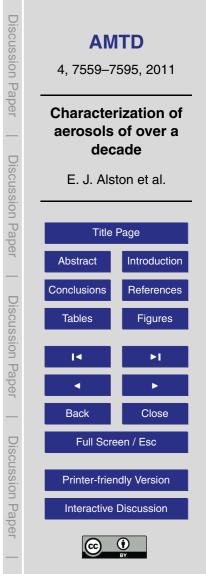
This study examines how aerosols measured from the ground and space over the US Southeast change temporally over a regional scale during the past decade. PM_{2.5} data consist of two datasets that represent the measurements that are used for regulatory

- purposes by the US EPA and continuous measurements used for quickly disseminating 5 air quality information. AOD data comes from three NASA sensors: the MODIS sensors onboard Terra and Aqua satellites and the MISR sensor onboard the Terra satellite. We analyze all available data over the state of Georgia from 2000-2009 of both types of aerosol data. The analysis reveals that during the summer the large metropolitan area
- of Atlanta has average $PM_{2.5}$ concentrations that are 50% more than the remainder 10 of the state. Strong seasonality is detected in both the AOD and $PM_{2.5}$ datasets; as evidenced by a threefold increase of AOD from mean winter values to mean summer values, and the increase in PM_{2.5} concentrations is almost twofold from over the same period. Additionally, there is good agreement between MODIS and MISR onboard the
- Terra satellite during the spring and summer having correlation coefficients of 0.64 and 15 0.71, respectively. Monthly anomalies were used to determine the presence of a trend in all considered aerosol datasets. We found negative linear trends in both the monthly AOD anomalies from MODIS onboard Terra and the PM_{2.5} datasets, which are statistically significant for $\alpha = 0.05$. Decreasing trends were also found for MISR onboard Terra and MODIS onboard Agua, but those trends were not statistically significant.

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Introduction 1

Over the past fifty or so years global ground-based measurements of solar radiation reaching the surface have shown first a decrease (i.e., dimming) and in the last fifteen years have shown an increase (i.e., brightening) Alpert et al. (2005); Gilgen et al. (2009); Wild et al. (2009). The solar dimming relates to increases in aerosol concentration that prevent incoming solar radiation from reaching the surface. During the 1980s, many industrialized countries enacted policies for controlling emissions of aerosols

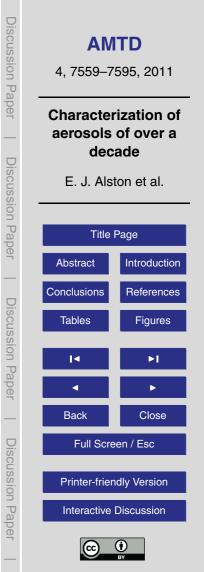


and their precursors. In the 1990s, a shift from dimming to brightening was reported at some locations (Streets et al., 2009). It is hypothesized that the magnitude of global warming has been masked due to solar dimming (Schwartz et al., 2010; Wild et al., 2009), thus linking this phenomenon to current climate.

- ⁵ Dutton et al. (2006) analyzed twenty seven years of NOAA/GMD surface solar irradiance data from five remote sites (Barrow, AK USA, Boulder, CO USA, Mauna Loa, HA USA American Somoa, and the South Pole) and concluded that while the sites span a large geographic area, the behavior of surface solar irradiance was similar (decreasing then increasing with time) across the sites. Wild et al. (2009) provide updates of
- ¹⁰ surface radiation measurements through 2005 and present evidence that brightening across large areas is ongoing and that anthropogenic contributions are an important factor in this phenomena. Streets et al. (2009) use model-predicted aerosol optical depth (AOD) to determine the regional nature of solar dimming/brightening. Their results indicate that the US, Europe and Russia have decreasing AOD values over a twenty-five year (1980–2005) period, and these regions also have a strong linear rela
 - tionship between AOD and surface radiation.

The regionality associated with this solar dimming/brightening is more nuanced when observed from a smaller regional perspective, e.g., regions of the US. For example, some studies found that that solar dimming/brightening is likely dominated by emissions from large urban areas (Alpert and Kischa, 2008; Alpert et al., 2005). Recent

- 20 sions from large urban areas (Alpert and Kischa, 2008; Alpert et al., 2005). Recent work by Wild et al. (2009) further substantiates this point by investigation of trends of solar dimming/brightening at multiple locations. The locations from the US all show a similar behavior, but each sites trend slope is different owing to the influence of differing aerosol mixtures and loading associated with each sites respective region. Long
- et al. (2009) investigated brightening of downwelling shortwave radiation at multiple US locations and found that collectively the brightening is significant, but that the varying degrees of brightening amongst the different sites suggested that research into dimming/brightening should address local to regional scales. Ultimately, understanding of dimming/brightening variations requires knowledge of spatiotemporal changes



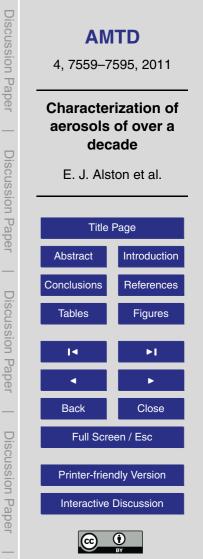
in aerosols on a regional basis. The focus of the present study is on a regional aerosol signal in the Southeast US.

This region is of interest because of the distinct aerosol mixtures associated with this geographic region that has been studied from the ground, yet little research has been done incorporating satellite data into longer term studies. There have been, for instance, large scale ground based measurement studies in the region consisting of a US Environmental Protection Agency (EPA) Supersite Study (Solomon et al., 2003), and ongoing work through the Southeastern Aerosol Research and Characterization Study (SEARCH) (http://www.atmospheric-research.com/studies/SEARCH/index.html). Ac-

- ¹⁰ cording to the EPA Our Nation's Air publication (http://www.epa.gov/airtrends/2010/ index.html), PM_{2.5} (particulate matter with an aerodynamic diameter less than 2.5 micrometers) concentrations nationwide have decreased by 19% since 2000, with the Southeastern US showing a decrease in annual PM_{2.5} concentrations. This region is distinctly characterized by aerosols composed of primarily sulfates and organics
- (Edgerton et al., 2005; Weber et al., 2007). Measurements and modeling studies have shown that organic aerosol formed by secondary processes is biogenic in origin and fairly homogeneous over the region (Lee et al., 2010). Goldstein et al. (2009) hypothesize that the haze commonly seen in the region during the warmer months is formed from secondary organic aerosols (SOA) formed from biogenic vilotile organic
- ²⁰ compounds (BVOC) that tend to cause a cooling radiative effect at the top of the atmosphere (TOA) and that these BVOC SOA aerosols form a layer aloft in the atmosphere with a dominant contribution to AOD during the summer.

Ground based measurements can provide high temporal concentration data over an extended period of time, yet these measurements are generally limited in their geo-

graphic coverage. A majority of ground based measurement sites are mostly in areas with high population densities. Additionally, ground based measurements are at best representative of aerosols in the lower atmosphere, mainly in the planetary boundary layer (PBL); as such, these measurements miss aerosols aloft, especially transport events. While satellites have an advantage in that they can view regions as a whole,

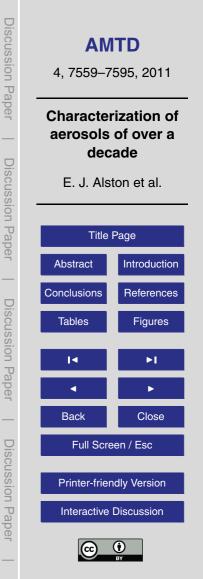


which enables better understanding of the regional aerosol dynamics, yet the satellites are limited temporally with only one or two sunlit overpasses of a region per day. The National Aeronautics and Space Administration (NASA) has launched a couple of satellites (Terra and Aqua) that have near global coverage daily with direct applica-

- tion to aerosols. The MODIS and MISR sensors onboard Terra have been operational for over ten years. MODIS onboard Aqua was launched in 2002. Each sensor provides a continuous timeseries of the aerosol optical depth (AOD) that can be directly related to the dimming/brightening phenomena (Hinkelman et al., 2009; Mishchenko et al., 2007). However, retrievals of AOD (a unitless measure of the amount of light at-
- tenuation over a set distance over the entire atmospheric column) are associated with a number of problems, especially over land such as deserts or urban environments. Liu and Mishchenko (2008) found that MODIS and MISR retrievals can disagree on a regional basis; yet, (Kahn et al., 2009, 2011) attempt to disprove those findings in concluding that MISR and MODIS retrievals are in agreement and provide details on the causes of the discrepancies between the two.

There have been studies that directly relate satellite AOD to PM_{2.5} concentrations in the US and the Southeastern US specifically (Engel-Cox et al., 2004; Gupta and Christopher, 2008, 2009; Zhang et al., 2009) for the purposes of predicting air quality. Our earlier work, used probabilities of AOD (based upon the linear regression relationship between AOD and PM_{2.5}) associated with different air quality index values as a means of prescribing air quality (Alston et al., 2011). Given the regional nature of aerosols and inherent difficulties and limitations in both satellite and ground based observations, it is important to utilize multiple sensors in aerosol analysis in order to develop as accurate understanding of aerosol behavior as possible especially when

viewed over a longer time period. A review about the state of the field with regards to aerosol characterization from space, suggests that different approaches other than relating PM_{2.5} and AOD through linear regressions are needed with respect to understanding the nature of these two aerosol measurements in future studies (Hoff and Christopher, 2009).



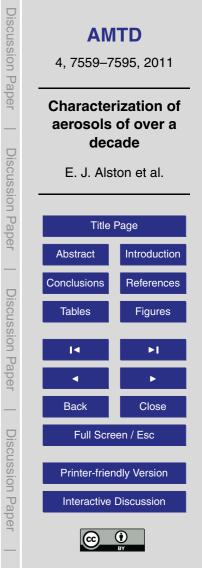
This is a motivation behind this study's goal to characterize aerosols in the US Southeast through analysis of ground and space based measurements from 2000–2009, with the emphasis on seasonal and interannual aerosol variations and understand these variations in the context of the radiative impact of aerosols on climate. The

- specific objectives are to examine the temporal changes of ground based PM_{2.5} and AODs from MODIS and MISR over the past ten years, determine common features and differences between these data records, determine if there is a discernible trend, and if a trend is present, what are the implications for the region in the context of the dimming/brightening phenomena. We analyzed ten years of AOD from MODIS and
- ¹⁰ MISR onboard Terra and eight years for MODIS onboard Aqua over a 5°x 5°box that encompasses the state of Georgia. This analysis also uses ten years of filter-based PM_{2.5} data provided by the EPA, and all available data from Georgia-run continuous PM_{2.5} monitors. This paper is organized as follows. Section 2 introduces the data and methods used in this study. Section 3 presents the results, and Sect. 4 concludes with a summary and discussion.

2 Data and methodology

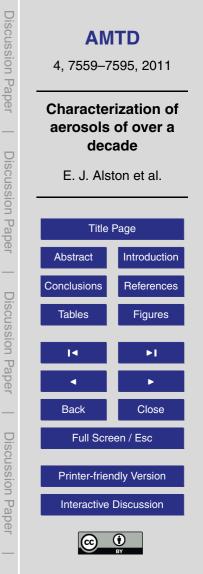
2.1 Ground-based PM_{2.5} data

We use surface PM_{2.5} measurements from two different networks: the ten years record from the national network of filter-based PM_{2.5} monitors courtesy of the EPA, and a seven year record of continuous PM_{2.5} measurements provided by the Georgia Dept. of Natural Resources. The location of the sites is shown in Fig. 1. The network operated by the Georgia Dept. of Natural Resources (http://www.air.dnr.state.ga.us/amp/) performs continuous hourly measurements using TEOMs (Tapered Element Oscillating Microbalance). Across Georgia, there are eighteen network sites located primarily within or near a city. For our study we use twelve sites. Seven of those sites are within the large metropolitan area of Atlanta and the remaining sites are smaller sized cities



and towns. Most of the stations have seven years of data; however, sites without at least five years of data were excluded from this analysis. Henceforth, this dataset will be annotated as $PM_{2.5,TEOM}$. The time coverage of this dataset is from 2003–2009.

- The second data set is provided by the EPA Air Quality Monitoring System (http: //www.epa.gov/airexplorer/index.htm). The data from this network are used for air quality regulatory purposes, e.g., attainment/non-attainment designations. Each monitor is a filter-based according to EPA-defined reference methods described in 40 CFR Part 53 (http://www.epa.gov/ttnamti1/40cfr53.html), and they must meet high quality control measures. Due to high level of quality control, there is usually a time lag from the measurement, the analysis, and finally making the data publicly available. Each station
- serves a different purpose; as such there are different repeat cycles. Population exposure monitors have daily concentrations; while the majority of sites have a 3-day repeat cycle. Monitors that capture background conditions have a 6-day repeat cycle. Similar to our methodology for the PM_{2.5,TEOM} dataset, we only use EPA sites within Georgia state lines, subsequently we use data from 29 sites. This dataset will be annotated as
- PM_{2.5,FRM}. Over half of the PM_{2.5,FRM} stations have data that encompass 2000–2009. To separate out Atlantas influence from the remainder of the state, we split each PM_{2.5} dataset into subsets depending on the geographical location of the considered sites. Ultimately, we have three subsets for each PM_{2.5}. We calculate a statewide
 mean for the *All GA* subset. The *Atlanta* subset is the mean exclusively using Atlanta sites. The last subset *Outside Atlanta* uses sites outside Atlanta for the calculated mean. For the PM_{2.5,TEOM} datasets, hourly means are averaged to create daily means. Those daily means are then used in subsequent analyses. Given the repeat cycle associated with the PM_{2.5,FRM} dataset, fill values were used to fill-in the gaps in the data record where measurements were not taken. Those complete timeseries were used in subsequent analyses.



2.2 MODIS data

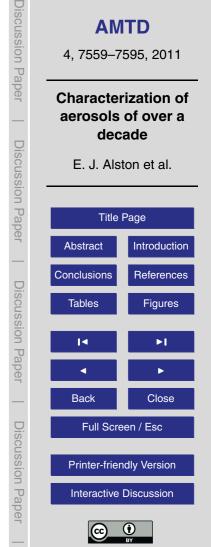
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The MODerate Resolution Imaging Spectroradiometer (MODIS) instrument flies onboard two NASA Earth Observing System (EOS) satellites: Terra and Aqua. Both sensors have near global coverage daily. Terra flies in the descending polar orbit with an equatorial crossing time of approximately 10:30 a.m., while Aqua flies in the ascending polar orbit with an equatorial crossing time of approximately 1:30 p.m. Generally.

- the satellites have overpass times over Georgia 5–15 min after their equatorial crossing times.
- MODIS passively measures reflected radiances from Earth across a broad wave length spectrum. It primarily uses three channels (0.47, 0.66, and 2.12 μm) to measure atmospheric aerosols over land (Levy et al., 2007). MODIS data are obtained from NASA LAADS (Level 1 and Atmosphere Archive and Distribution System). The analysis is performed with MODIS Collection 5 Level 2 data, which have a nominal resolution of 10 × 10 km² at nadir. The variable of most importance to this study is *Optical Depth* Land and Ocean at the 550 nm wavelength, which incorporates only the highest quality
- retrievals.

2.3 MISR data

The Multi-angle Imaging SpectroRadiometer (MISR) flies onboard of the Terra satellite together with MODIS. MISR is a multi-angle imaging instrument consisting of nine
cameras with view angles of ±70.5°, ±60.0°, ±45.6°, ±26.1°, and 0°(nadir), operating in four spectral bands centered at 446 nm (blue), 558 nm (green), 672 nm (red), and 867 nm (near infrared). In global observing mode, the spatial resolution of the red band is 275 m in all nine cameras, the other bands are re-sampled to 1.1 km resolution in all the cameras, except the nadir, which preserves the full 275 m resolution
in all four bands. The common swath width is ≈400 km and global coverage is obtained every nine days at the equator and more frequently at higher latitudes (Diner



resolution, and particle size, shape, and single-scattering albedo are retrieved in addition to AOD (Martonchik et al., 2002, 2009) A global comparison of coincident MISR and AERONET sunphotometer data showed that overall about 70 % to 75 % of MISR AOD retrievals fall within 0.05 or 20 % of AOD, and about 50 % to 55 % are within 0.03

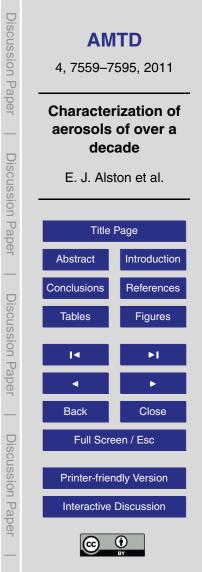
- or 10% of AOD, except at sites where dust or mixed dust and smoke are commonly found (Kahn et al., 2010). MISR data were obtained from NASA Langley ASDC (Atmospheric Science Data Center). The analysis is performed with MISR version 22 Level 2 aerosol data. The used AOD values are "best estimate AOD" at MISR green (558 nm) band that combines the land and ocean AOD products.
- For each satellite, we create a subset based on the latitude/longitude box 30° N– 35° N and 80° W–85° W. All the satellite pixels contained within that latitude/longitude box are averaged together to create a regional mean AOD value on a daily basis for each satellite sensor. The daily mean AOD values are used in the subsequent analyses. For spatial analysis the nominal Level 2 products are used to create maps of AOD from both Terra instruments. The daily granules are averaged on a global grid (0.25° × 0.25° for MODIS and 0.2° × 0.2° for MISR). These grids are then averaged to create seasonal means of AOD fields for the ten year time period covering the afore-

3 Results

20 3.1 Seasonal cycle

mentioned latitude/longitude box.

Where available, we analyzed 10 yr of $PM_{2.5,TEOM}$, $PM_{2.5,FRM}$, and AOD data from MODIS Terra and Aqua, and MISR Terra to investigate the seasonal aerosol signatures over the US Southeast. Considering only spring and summer seasons in our previous study (Alston et al., 2011), we found that $PM_{2.5}$ and AOD have different seasonal traits with AOD values almost doubling during the summer compared to values in



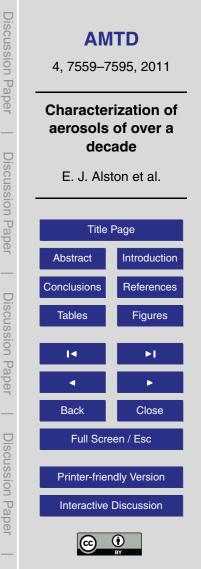
the spring. Here we calculated 10-yr (if available) averages of each month for both the satellite and $PM_{2.5}$ datasets. The results are shown in Fig. 2. In analyzing a full calendar year instead of just spring and summer, here we determine that summer (June–August) AOD (0.32–0.35) is almost tripled from wintertime (December–February) AOD

- (0.08–0.1). MODIS Terra has the highest average AOD. Generally speaking, both Terra sensors (MODIS and MISR) have higher AOD than MODIS Aqua. During the summer months the difference between the MODIS AOD sensors and MISR AOD is about 0.1. While the difference between the MODIS AOD sensors at its highest is about 0.025. The noted 3x increase cannot be fully attributable to PM_{2.5} increases over the
- ¹⁰ same period. The different PM_{2.5} datasets behave differently, with PM_{2.5,TEOM} doubling concentrations during the summer whereas PM_{2.5,FRM} shows only a modest increase over the same period ($\approx 10.0 \,\mu g \,m^{-3}$ during the winter to $\approx 18.0 \,\mu g \,m^{-3}$ during the summer). It is possible that some of the differences seen between the datasets are that the PM_{2.5,TEOM} dataset only has 7 yr of data compared with 10 yr of the EPA dataset.
- ¹⁵ Another possibility could be due to the differences in measurement techniques used. The standard error (standard deviation/number of observations) of the means of both datasets show more variability during the warmer months, see Fig. 2.

Timeseries of monthly mean data for each year are shown in Fig. 3. The winter months have the lowest values of AOD and $\rm PM_{2.5}$, while the summer months have the

- ²⁰ highest. Specifically, July and August have the highest AOD values with maximums over the years varying between 0.5–1.5, with January and December having the lowest values between 0.2–0.55. MODIS Aqua has a much tighter AOD envelope with winter-time AOD values between 0.05–0.08 and summertime AOD values between 0.25–0.5. The year 2007 has anomalously high values in all the datasets. For a majority of the year, the satellite datasets have small amounts of interannual seasonal variability, with
- the highest amounts of interannual variability occurring in the summer.

The $PM_{2.5}$ datasets have more interannual variability than the satellite datasets. Breaking the $PM_{2.5}$ datasets into different geographic regions allows us to evaluate the effect of the large urban area of Atlanta on the region as a whole. Atlanta



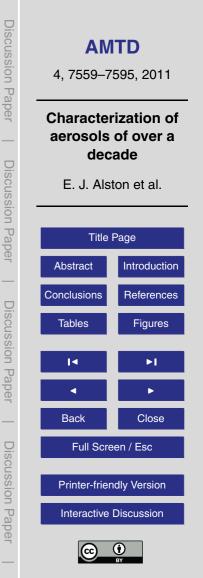
concentrations from both PM_{2.5,TEOM} and PM_{2.5,FRM} have more variability throughout the year when compared to stations outside the Atlanta metropolitan area, see Fig. 3d– i. Our results suggest that during the summer there is a complex dynamic relationship between regional background PM_{2.5} concentrations and anthropogenic emissions that lead to Atlanta having a 50 % or more increase in surface concentrations that are not

observed elsewhere in the state.

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Our analysis to this point has shown relatively good agreement between the satellite sensors. To further investigate the agreement between sensors, specifically MODIS Terra and MISR Terra, we analyze seasonal coincident AOD monthly means. We define each season in standard fashion: winter (December, January and February), spring

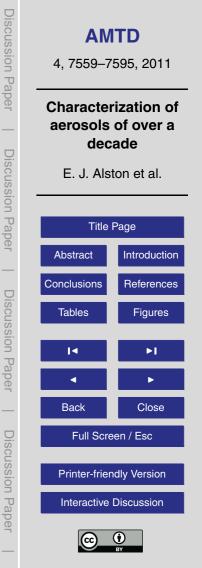
- (March, April and May), summer (June, July and August), and fall (September, October and November). Figure 4 presents the comparison of AODs. In most seasons, MODIS reports higher AOD values than MISR. Although our results from Fig. 3 show that on a monthly basis the differences between the two sensors are smallest during
- the fall and winter, our seasonal analysis (Fig. 4) shows more variance between the sensors. Not surprisingly, the linear regression slopes (0.33 for winter and 0.54 for fall) are not close to 1, and the subsequent correlation coefficients are 0.33 and 0.57, respectively. The outliers in scatterplots are possibly due to retrieval biases, differences within the retrieval algorithms, and cloud effects. Remer et al. (2008) found that on
- ²⁰ a global scale, aerosols near clouds only occur less than 1 % of the time over land; however, they note that AOD values near clouds can double the reported AOD due to subpixel cloud contamination (Zhang et al., 2005), 3-D effects (Wen et al., 2007), and increase of AOD due to increased humidity near clouds (Koren et al., 2007). (Kahn et al., 2009) found that MODIS AOD values are lower than MISR AOD values for AOD
- ²⁵ below 0.2, which could be related to Collection 5 algorithm changes that allow for negative AOD retrievals. The spring and summer seasons produce the greatest agreement between the two sensors with correlation coefficients of 0.64 and 0.71, respectively. Hygroscopic growth of aerosols due to higher relative humidity in the summer also possibly influence the agreement between the sensors. An additional influence could



be the weather pattern dynamics with the spring and summer seasons experiencing large-scale high pressure systems that can persist, which likely results in increased AOD values for both sensors despite their differences in viewing geometry. Interestingly, the signs of the y-intercepts are negative for spring and summer seasons. Pos-

- sible explanations for this include that we do not force our linear regressions through zero, and due to systematic underestimation of AOD by MISR (Kahn et al., 2009), the regression line is pulled downward. Nevertheless, our results suggest good agreement between the two sensors over the past ten years. Yet Liu and Mishchenko (2008) reported larger disparities between the two. We should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishchenko, 2000) and a statement of the should note that (Liu and Mishche
- 2008) only consider two months (January and July) from 2006, and they consider a region (Eastern US) that is spatially larger than our area and contains multiple sources of aerosols (e.g., large metropolitan area). While our study region only contains one large metropolitan area, i.e., Atlanta.

Figure 5 shows how the seasonal means for each dataset change over time. As expected, spring and summer seasons show the most variance over the years. For 15 instance, in the year 2000 MODIS Terra and MISR Terra had summer AOD means of 0.38 and 0.29, respectively, and by 2009 the means were 0.21 and 0.24. Also, even though our considered spatial domain is relatively small (5°by 5°), our seasonal means are similar in behavior to those of East North America as shown in Remer et al. (2008) where Level 3 1° × 1° globally gridded AOD are used for regional seasonal anal-20 ysis, yet our seasonal means are higher. In the PM2.5 datasets there appear to be different behaviors. The PM_{2.5.FRM} values all appear to be decreasing with time. In 2000, $PM_{2.5 EBM}$ concentrations were around 22 µg m⁻³, but by the end of the decade they had decreased to around $14 \mu g m^{-3}$. The spring, fall, and winter seasons have similar behaviors, with summer being the exception. The three seasons also show 25 similar behavior across all of Georgia, yet during the summer our results suggest that Atlanta is dominating concentrations across the state. The difference between the All GA and Atlanta means at most varied around $2 \mu g m^{-3}$, and there is a larger difference (4 μ g m⁻³) between the All GA means and the Outside Atlanta means. Alston



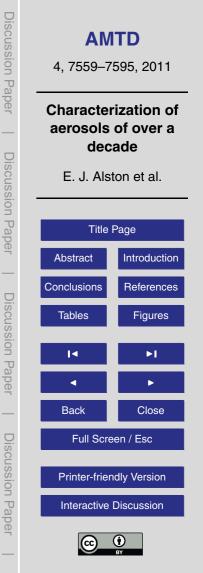
et al. (2011) highlighted how the spring of 2007 was anomalous in both AOD and $PM_{2.5}$ concentrations compared with other springs due to the large wildfire that burned for almost two months. It is likely that if the wildfire had not occurred, the spring means would decrease with time. Despite increased means for 2006 and 2007, the $PM_{2.5,TEOM}$ dataset appears to generally decrease with time.

5

The aerosol seasonality was also examined through an analysis of satellite AOD fields over the past 10 yr. In particular, we were interested in understanding if there are any discernible AOD differences from the large metropolitan area of Atlanta and the remainder of the state. Seasonal maps of AOD from MODIS Terra and MISR Terra are

- ¹⁰ shown in Fig. 6: winter mean AOD (a and d), summer mean AOD (b and e), and the difference between the two seasons in (c and f). These maps, specifically the seasonal difference maps provide comparison to similar figures in Goldstein et al. (2009), see their Fig.1. Our spatial analysis does not strongly resemble the features seen in Goldstein et al., namely the large area of AOD (AOD > 0.25) over the broader Southeastern
- US. It should be noted that a major difference between this study and theirs is that we use a finer resolution product (Level 2) which is gridded to finer resolution grid than is provided by the Level 3 (1° × 1°) monthly mean product used by Goldstein et al. The Level 3 products produces smoother appearing maps that can likely mask large point sources (e.g., industrialization, large metropolitan areas). This study also uses data
 from 2000–2009, whereas their study encompassed 2000–2007.

The MODIS maps suggest that the Atlanta area has slightly higher AOD from the remainder of the region, see Fig. 6a–c. The MISR maps do not appear to capture the AOD signal in Atlanta as well as MODIS, see Fig. 6d–f; however, both sensors show very low AOD during the winter season with AOD values <0.1, though there are some ²⁵ areas near the coastlines and over the ocean where AOD > 0.1 (Chu et al., 2002; Kahn et al., 2007; Levy et al., 2005). The summer season presents a more varied spatial representation. As noted early, there is almost a 3x increase from winter AOD values. One common feature between the sensors is that the background region (the region minus Atlanta) appears fairly uniform in AOD. The difference plots (Fig. 1c and f



suggest variation across the region that is not seen in Fig. 1 of Goldstein et al. (2009). In summary, the spatial analysis presented here shows slight differences between Atlanta and the remainder of the region at least from the MODIS Terra perspective, and this analysis shows more marked seasonality in spatial extent and magnitude than pre-

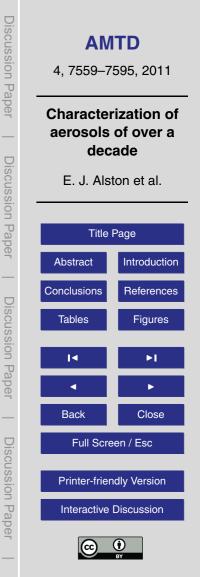
viously shown by Goldstein et al. (2009). Finer scale spatial resolution of satellites will likely aid the differentiation of urban centers from background conditions. Until newer satellite sensors are available with finer resolution, regional scale analysis are likely to remain the current standard, which has implications for air quality forecasts that want to incorporate satellite data into these forecasts on a state or smaller scale.

10 3.2 Interannual variability and trends

To examine interannual variability of aerosol in the Southeast US, we analyzed monthly means of satellites AOD and ground based $PM_{2.5}$ data, including analyses of anomalies and trends. Figure 7 presents the timeseries of monthly mean AODs for MODIS Terra, MISR Terra, and MODIS Aqua, along with timeseries of monthly mean

- ¹⁵ PM_{2.5} concentrations for the two ground datasets. When viewed over the past ten years, the satellites have generally good agreement with each other. Though there are differences in AOD magnitudes between MODIS Terra and MISR Terra, their behavior over time is quite similar. The difference between minima (\approx 0.1) and maxima (\approx 0.4) for the MODIS sensors is about 0.3. Another way to put that is according to the MODIS
- ²⁰ sensor, AOD almost quadruples from the lowest values in winter to the highest values in summer. MISR appears to have quite dramatic fluctuations as well, with its minima ≈ 0.3 and its maxima ≈ 0.8 . The interannual variability makes it difficult to determine if there is a trend over time.

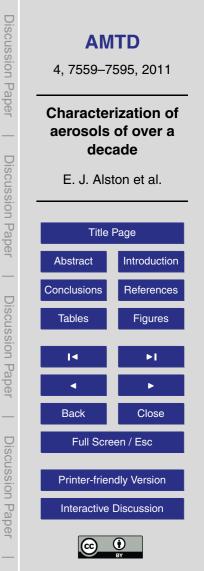
In contrast, the $PM_{2.5}$ datasets show a distinctly decreasing trend over time. Both the maxima and minima for these datasets have decreased by 5–8 µg m⁻³. The seasonality is present in the $PM_{2.5}$ datasets, but not as pronounced as the AOD datasets. When viewed together (both AOD and $PM_{2.5}$ datasets), the peaks and valleys in



the timeseries correspond well together. For instance, the correlation coefficient of MODIS Terra vs. PM_{2.5,FRM} *All GA* and MISR Terra vs. PM_{2.5,FRM} *All GA* is 0.72 and 0.73, respectively; however, the correlation coefficient of MODIS Aqua vs. PM_{2.5,FRM} *All GA* is 0.8. Correlation analysis between the satellites and PM_{2.5,TEOM} *All GA* yields 0.84, 0.81
and 0.84, respectively. Finally, Fig. 8 presents AOD and PM_{2.5} concentrations over the past 10 yr in terms of yearly means calculated from monthly means. It is readily apparent that there is a decreasing trend across all datasets. One point of incongruity occurs in 2007. In the AOD datasets, 2007 is high compared with years 2006 and 2008, yet 2007 does not have this peak in the PM_{2.5} dataset. Alston et al. (2011) suggested that aerosols aloft associated with aerosol transport of local and long range haze and biomass burning events could be a likely explanation.

As shown in Fig. 7 there is strong seasonality, which makes the determination of any increasing/decreasing trend difficult. The first step in the determination of a trend is to fit the timeseries with a linear regression. The second step is to access if the slope is

- statistically different from zero by using t-test for $\alpha = 0.05$. Though there was no trend easily detected in Fig. 7, we fit each satellite AOD with a linear regression and determined that all the datasets did not have a statistically significant slope. As mentioned earlier, the PM_{2.5} datasets appear to be decreasing with time. The linear regression for PM_{2.5,FRM} all have slopes that are significant for $\alpha = 0.05$. In other words, the detected
- decrease in the timeseries is valid with some certainty. The PM_{2.5,TEOM} datasets have more varied results. The slopes for PM_{2.5,TEOM} All GA and PM_{2.5,TEOM} Atlanta are not statistically significant, yet PM_{2.5,TEOM} Outside Atlanta is significant. Our previous results suggest that metropolitan area of Atlanta concentrations likely skew the statewide average towards higher values due to the majority of the TEOM monitors (7 or 60%)
- ²⁵ being in the metropolitan area of Atlanta area. Our results also hint that the rest of the state is indeed experiencing decreasing $PM_{2.5}$ concentrations, but the anthropogenic emissions especially in the summer in the metropolitan area of Atlanta are likely masking this decreasing trend. Another possible explanation for why one $PM_{2.5}$ dataset shows a decreasing trend and the other does not is the difference in length of the data



records. We believe that given the air quality control policies in place, the $PM_{2.5,TEOM}$ dataset will likely show a statistically decreasing trend given more time.

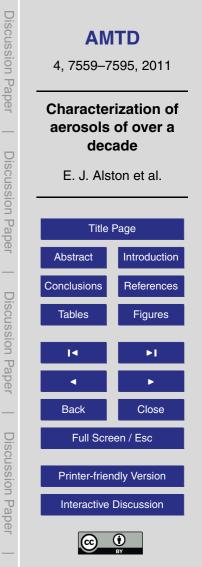
Ultimately, it is necessary to remove the seasonal signal in order to access the presence of any true trends. We calculate a ten-year mean of every month, and subtract ⁵ each month from the 10-yr mean of that month. For example, if the ten-year January average is 0.18, then 0.18 is subtracted from each January in the dataset, thus we are using anomalies from the 10-yr monthly mean to detect trends over the past 10 yr. The resulting timeseries of anomalies for both satellite and PM_{2.5} datasets are shown in Fig. 9a-c. The anomaly timeseries are fit with linear regressions to determine the trend and are shown by the dashed line in Fig. 8. MODIS Terra was the only satellite dataset 10 to have a statistically significant slope. We believe the smaller range of MISR AOD is why that dataset does not have a significant slope, while the MODIS Agua dataset is only 8 yr long. It is possible that as time progresses the MODIS Aqua dataset will show a decreasing trend with certainty. By removing the seasonal component within the PM_{2.5} datasets revealed statistically significant decreasing trends, see Fig. 9b–c. The 15 linear regression variables (slope and y-intercept) are summarized for each dataset in Table 1. The regression variables are calculated on a per decade basis. We hypothesize that removing the strong seasonality from those datasets the summertime peaks in concentrations were minimized thus allowing a true and statistically significant trend

²⁰ to emerge.

4 Conclusions and discussion

We analyzed aerosol data from both ground based ($PM_{2.5}$) and space based (satellite AOD) platforms to examine the seasonality and interannual variations of the regional aerosol signal, and to detect if there was any discernable trends over the past ten years.

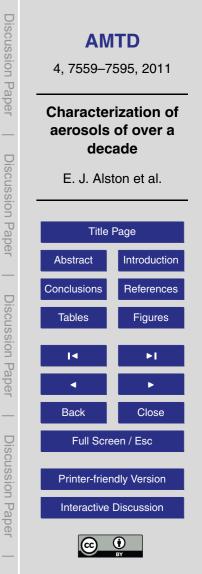
²⁵ We found that strong seasonality exists in both the AOD and PM_{2.5} datasets where mean summertime AOD is nearly three times higher than mean wintertime AOD, and



mean summertime $PM_{2.5}$ concentrations are almost twice as high as mean wintertime concentrations. Another factor that possibly influences the seasonality is the effect of hygroscopic aerosol growth during the summer months, given higher relative humidity in the summer. Though satellite retrieval algorithms do not directly incorporate relative

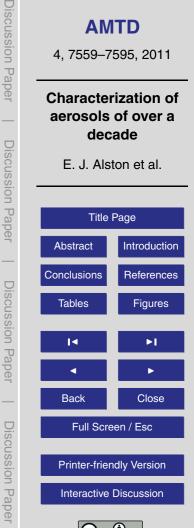
- ⁵ humidity, the retrievals are affected (Wang and Martin, 2007). Additionally over the past ten years, the PM_{2.5} dataset used for regulatory purposes (PM_{2.5,FRM}) agree quite well with the satellite AOD measurements of aerosols. The correlation coefficients between PM_{2.5,FRM} and AODs from MODIS Terra are 0.72, for MISR Terra are 0.73, and for MODIS Aqua are 0.8.
- ¹⁰ We found that MODIS onboard Terra and MISR onboard Terra agree well with each other during the warmer months with correlation coefficients of 0.67 for spring and 0.71 for summer. It is possible that cloud cover and inherent differences in sensor sensitivity explain the reduced agreement during the cooler months. Trend analysis was performed to establish baselines of different aerosol measures. We use t-tests
- ¹⁵ of the slopes for α = 0.05 to determine whether the calculated slopes are statistically different from zero. Trend analysis of monthly means AOD revealed that none of the satellite datasets shows a statistically significant negative trend. Yet the PM_{2.5,FRM} FRM monthly mean timeseries does have statistically significant negative trends. Given the strong seasonality, we removed the seasonal component to create monthly mean
- ²⁰ anomalies. Trend analysis of the monthly mean anomalies yielded that MODIS onboard Terra has a statistically significant negative trend, and all the PM_{2.5} datasets have statistically significant negative trends. It should be noted that for MODIS onboard Terra, this detected trend could be impacted by degradation of the blue channel used in MODIS retrievals over land, yet even with this drift taken into account the re-
- trieved values are within the acceptable error envelope (Kahn et al., 2011; Levy et al., 2010).

Our results question the Goldstein et al. (2009) hypothesis on a dominant contribution of SOA from biogenic emission to AODs in the region. AOD is a columnaveraged measurement that cannot readily differentiate between sources without a



priori information. If the conversion rate between BVOC and SOA is primarily temperature driven, then if the temperature record was found to neither increase or decrease (Menne et al., 2009) then biogenic SOA is unlikely to be the driver behind the negative trends in AOD. Following this reasoning the primary driver behind the negative trend appears to be anthropogenic sources which are monitored and controlled through air quality policies. Another facet of the Goldstein et al. (2009) hypothesis is that these BVOC associated SOA are formed in an aerosol layer aloft and thus ground based sensors would not likely capture the additional aerosol load of those aerosols. One would not expect trends in the PM_{2.5} records, yet our results show decreasing seasonal and yearly trends. These results suggest that ground based monitors are measuring some

- 10 yearly trends. These results suggest that ground based monitors are measuring some portion of these SOA aerosols. Of course, this result requires additional measurement of aerosol profiles in this region for confirmation purposes. However, given the current state of measurement techniques it is not a simple exercise to differentiate between SOA of anthropogenic and biogenic sources (Weber et al., 2007). Additionally obser-
- vational evidence for a layer of SOA aloft as measured by aircraft field campaigns is not supported (Heald et al., 2011). Finally, the spatial analysis presented here somewhat agrees with that shown in Goldstein et al. (2009). Of significance is that our results are different in spatial features (not smooth continuous fields of AOD) and magnitude (the difference between summer and winter is higher).
- Our analysis suggests that air quality policies and controls placed upon PM_{2.5} precursors have resulted in appreciable decreases in aerosols in the US Southeast. Our results also suggest that this region is experiencing solar brightening associated with decreasing concentrations of aerosols. Ground based measurements of solar irradiance in the region would be necessary to confirm our conclusions. Currently,
- there is no such monitoring being done. Our analysis also provides a useful baseline for naturally derived aerosols representative of background conditions in this region of the US. Establishing the background helps to delineate the PM_{2.5} contributions of the metropolitan area of Atlanta. Thus, it is likely that future air quality control strategies will need to focus upon the anthropogenic component, while also incorporating



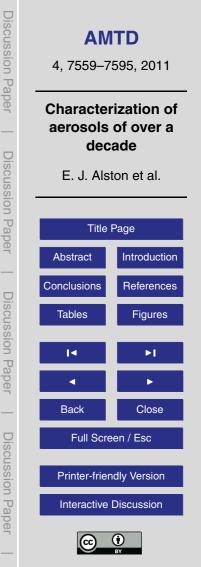
naturally occurring aerosols. Additionally, the air quality control policies that have likely resulted in solar brightening might have potential climatic trade-offs. As such these longer-term analyses are critical for evaluation of the air pollution regulatory policies, and these analyses can serves as baselines of measures that can be used to access

- impacts of future policies and climate change. The methodology applied here is readily applicable to regions that have sufficient ground based aerosol measurements so long as the chosen area is large enough for sufficient satellite coverage. The need for finer scale resolution satellite sensors will aid a host of applications seeking to do more detailed regional and local scale analyses. Users of satellite data need to be aware of
- possible bias within the data at land-water boundaries, which is an important consideration given that so many highly populated areas are near coasts. It is possible that with newer sensors better treatment of these issues will be addressed. Our future work will focus upon the climatic impacts of the decreasing aerosol trend on this region.

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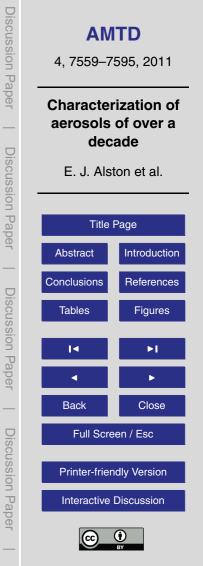
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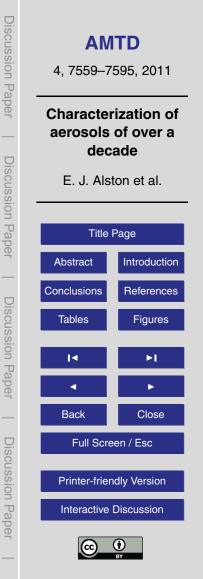
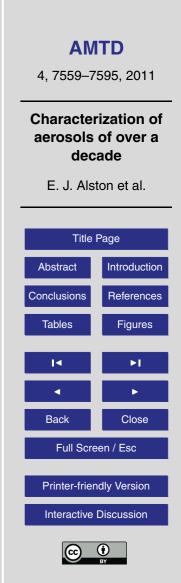


Table 1. Linear regression coefficients for satellite and $PM_{2.5}$ datasets. Significance for $\alpha = 0.5$ is denoted in bold. The units for slope are (AOD month⁻¹ or μ g m⁻³ month⁻¹), and the units for y-intercept are (AOD or μ g m⁻³).

Delevel	Slope		Y-intercept	
Dataset	With trend	Without trend	With trend	Without trend
MODIS Terra	-0.000415	-0.000434	0.214	0.025
MISR Terra	-0.000177	-0.000112	0.15	0.11
MODIS Aqua	-0.000219	-0.000275	0.196	0.17
PM _{2.5.FRM} AllGA	-0.46	-0.0448	17.75	2.781
PM _{2.5.FBM} Atlanta	-0.0488	-0.0472	18.187	2.951
PM _{2.5,FRM} OutsideAtlanta	-0.0301	-0.302	16.029	1.821
PM _{2.5.TEOM} AllGA	-0.0317	-0.0319	15.876	2.501
PM _{2.5.TEOM} Atlanta	-0.0324	-0.0319	16.325	2.561
PM _{2.5,TEOM} Outside Atlanta	-0.0335	-0.0385	15.876	2.696

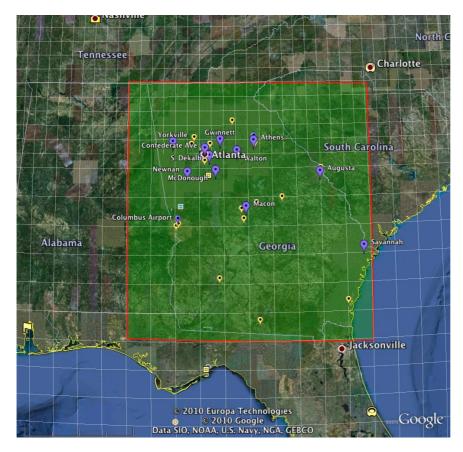


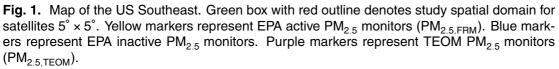
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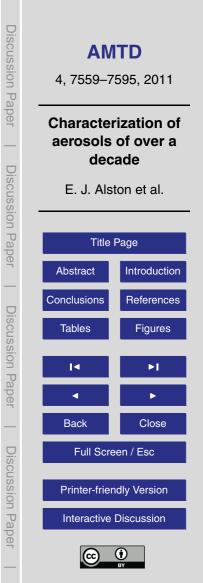
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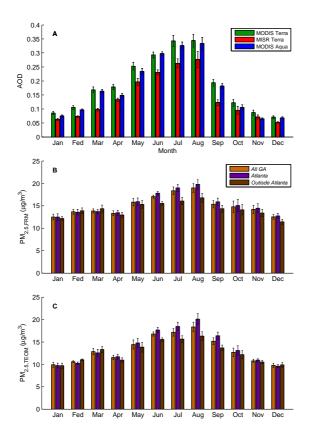
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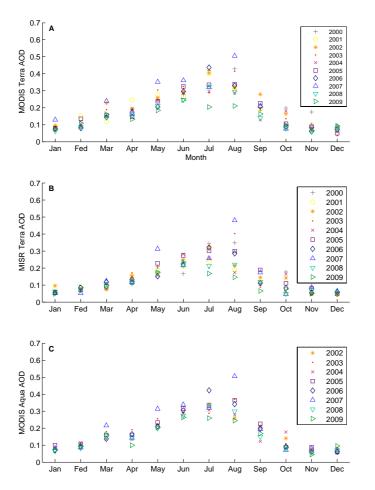


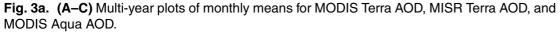


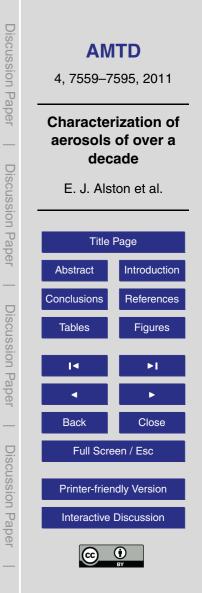


Discussion Paper **AMTD** 4, 7559-7595, 2011 Characterization of aerosols of over a decade **Discussion** Paper E. J. Alston et al. **Title Page** Introduction Abstract Conclusions References **Discussion** Paper Tables Figures Back Close **Discussion** Paper Full Screen / Esc Printer-friendly Version Interactive Discussion

Fig. 2. (A) Bar plots of ten-year means by month for MODIS Terra AOD, MISR Terra AOD and MODIS Aqua AOD. (B) Same as (A) except for $PM_{2.5,FRM}All GA$, $PM_{2.5,FRM}Atlanta$ and $PM_{2.5,FRM}Outside Atlanta$. (C) Same as (A) except for $PM_{2.5,TEOM}All GA$, $PM_{2.5,TEOM}Atlanta$ and $PM_{2.5,TEOM}Outside Atlanta$. The units for all $PM_{2.5}$ are μ g m⁻³. Whiskers represent \pm standard error of the mean for each respective dataset.







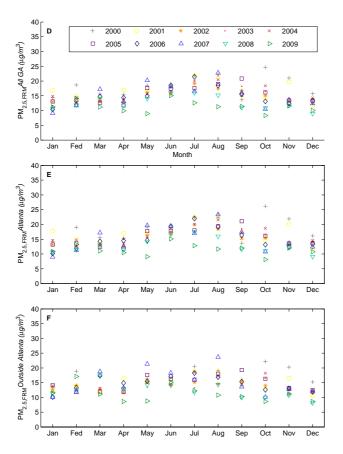
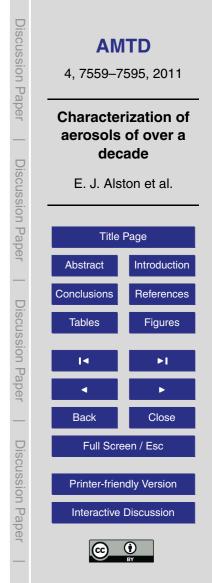


Fig. 3b. (D–F) Multi-year plots of monthly means for $PM_{2.5,FRM}All GA$, $PM_{2.5,FRM}Atlanta$, and $PM_{2.5,FRM}Outside Atlanta$. The units for all $PM_{2.5}$ are $\mu g m^{-3}$.



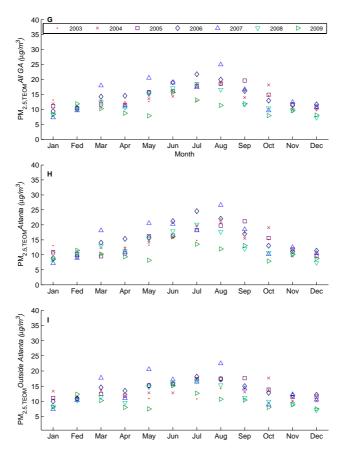
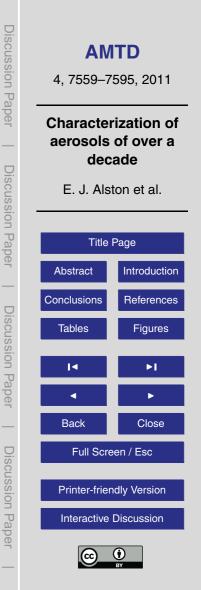
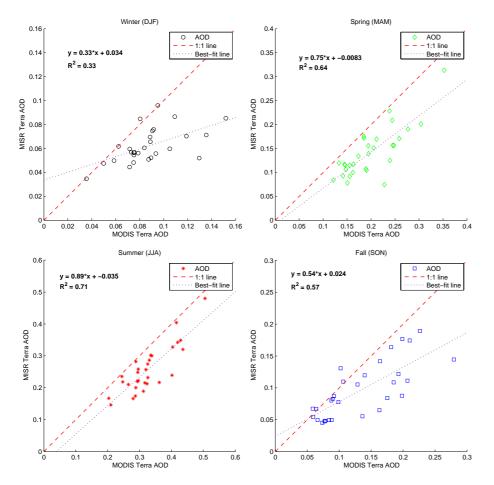
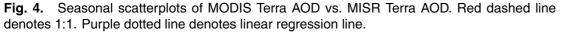
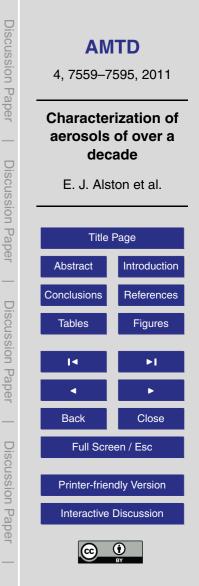


Fig. 3c. (G–I) Multi-year plots of monthly means for $PM_{2.5,TEOM}All GA$, $PM_{2.5,TEOM}Atlanta$, and $PM_{2.5,TEOM}Outside Atlanta$. The units for all $PM_{2.5}$ are $\mu g m^{-3}$.









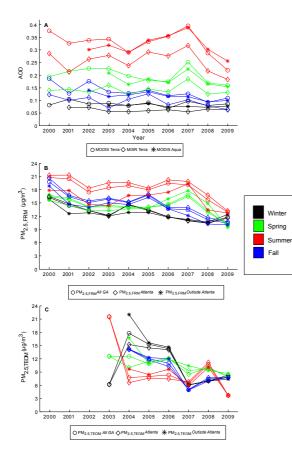
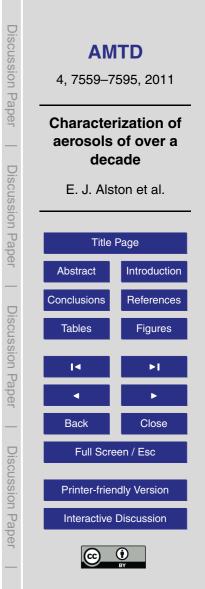


Fig. 5. (A) Timeseries of seasonal means for MODIS Terra AOD, MISR Terra AOD and MODIS Aqua AOD. (B) Same as (A) except for $PM_{2.5,FRM}AllGA$, $PM_{2.5,FRM}Atlanta$, and $PM_{2.5,FRM}Outside Atlanta$. (C) Same as (A) except for $PM_{2.5,TEOM}AllGA$, $PM_{2.5,TEOM}Atlanta$, and $PM_{2.5,TEOM}Outside Atlanta$. The units for all $PM_{2.5}$ are $\mu g m^{-3}$.



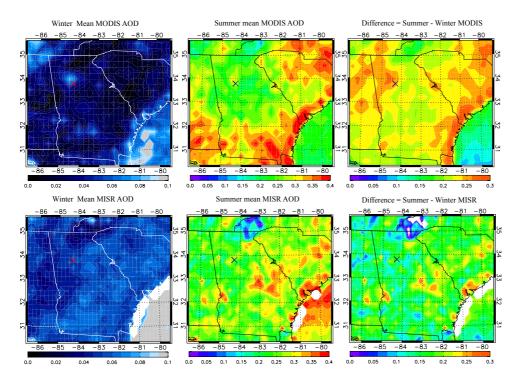
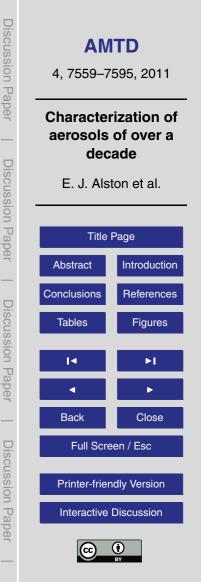
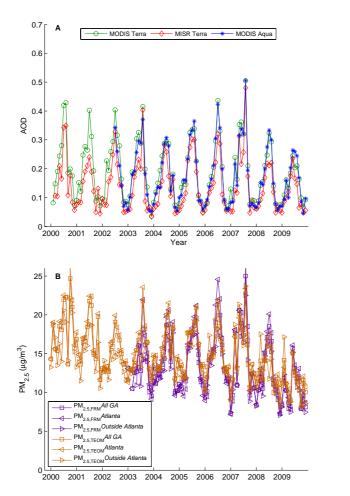
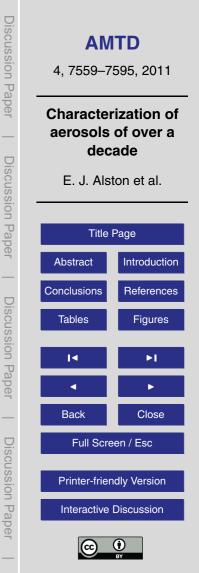


Fig. 6. Maps of satellite AOD. **(A)** Winter mean AOD for MODIS Terra. **(B)** Summer mean AOD for MODIS Terra. **(C)** Difference between summer mean AOD minus winter mean AOD for MODIS Terra. **(D)** Same as **(A)** except for MISR Terra. **(E)** Same as **(B)** except for MISR Terra. **(F)** Same as **(C)** except for MISR Terra. In **(A** and **D)** the red "X" denotes Atlanta, GA. In **(B–C** and **E–F)** the navy X denotes Atlanta, GA.









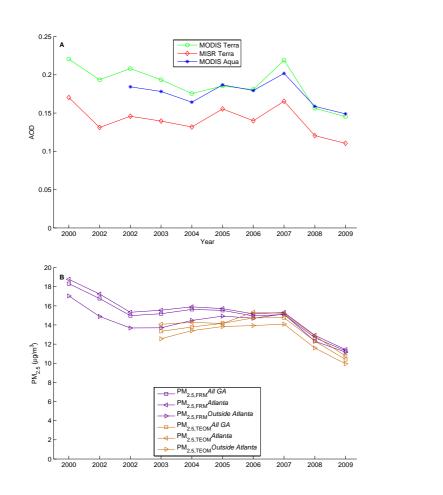
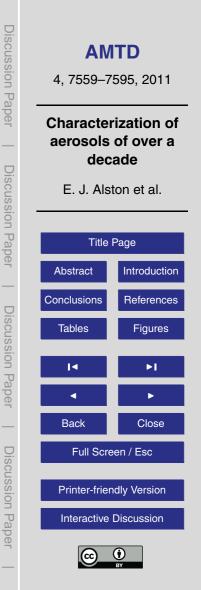
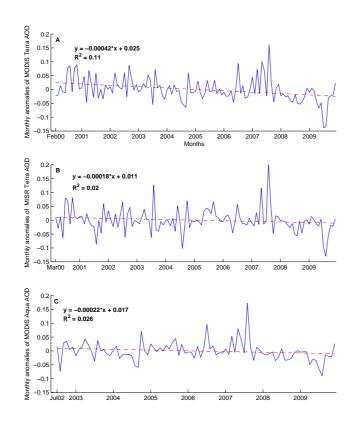
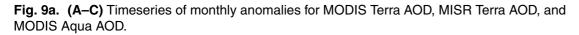
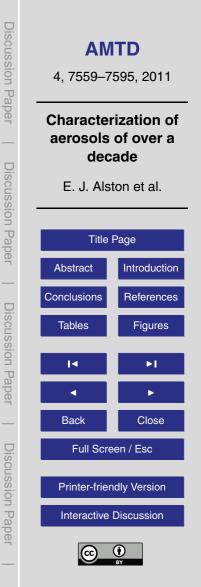


Fig. 8. Timeseries of multi-year means for satellite AOD and $PM_{2.5}$ datasets. The units for all $PM_{2.5}$ are μ g m⁻³.









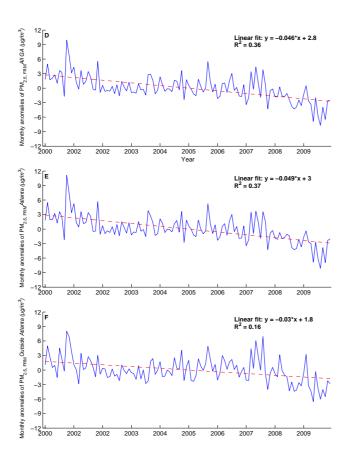
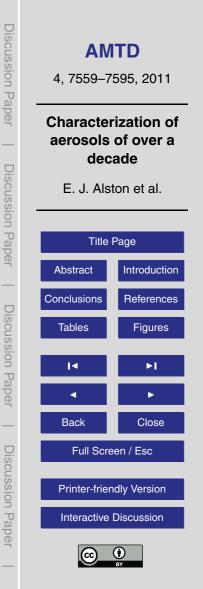


Fig. 9b. (D–F) Timeseries of monthly anomalies for $PM_{2.5,FRM}$ *All GA*, $PM_{2.5,FRM}$ *Atlanta*, and $PM_{2.5,FRM}$ *Outside Atlanta*. The units for all $PM_{2.5}$ are μ g m⁻³.



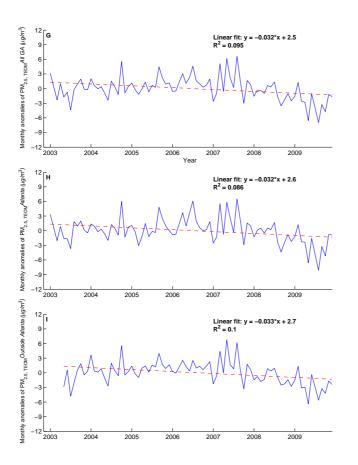


Fig. 9c. (G–I) Timeseries of monthly anomalies for $PM_{2.5,TEOM}All GA$, $PM_{2.5,TEOM}Atlanta$, and $PM_{2.5,TEOM}Outside Atlanta$. The units for all $PM_{2.5}$ are $\mu g m^{-3}$.

