



Seasonal and Spatial Changes in Trace Gases over Megacities from AURA TES Observations

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Abstract

The AURA TES instrument is collecting closely spaced observations over 19 megacities. The objective is to obtain measurements that will lead to better understanding of the processes affecting air quality in and around these cities, and better estimates of the seasonal and interannual variability. We explore the
20 TES measurements of ozone, ammonia, methanol and formic acid collected around the Mexico City Metropolitan Area and in the vicinity of Lagos (Nigeria). The TES data exhibit seasonal signals that are correlated with AIRS CO and MODIS AOD, with in situ measurements in the MCMA and with GEOS-Chem model output in the Lagos area. TES was able to detect an extreme pollution event in the MCMA on April 9, 2013, which is also evident in the in situ data. TES also shows that biomass burning
25 has a greater impact south of the city than in the caldera where Mexico City is located. TES measured enhanced values of the four species over the Gulf of Guinea south of Lagos. Since it observes many cities from the same platform, with the same instrument and applies the same retrieval algorithms, TES data provide a very useful tool for quickly comparing air quality measures of two or more cities. We compare the data from the MCMA and Lagos, and show that while the MCMA has occasional
30 extreme pollution events, Lagos consistently has far higher levels of these trace gases.

1 Introduction

The world's megacities, defined as urban agglomerations with a population over ten million (UN, 2016), currently house close to 500 million people, about one eighth of the global urban population



(UN, 2014). They have increased in number, from two in 1950 (New York-Newark and Tokyo-Yokohama), to 28 in 2014, and have become increasingly larger. This growth is expected to continue, as more and more people leave rural areas for better opportunities and a higher quality of life. Yet while it has been shown that urban dwellers are financially better off, better educated and better fed, this improvement does come at the cost of strained services and increased congestion, causing, among other problems, increases in urban water and air pollution. The fastest growing megacities in the past few decades have been located primarily in the developing world (UN, 2014), where rapid industrialization in combination with high population density and limited emission controls is leading to serious air quality problems (UN, 2014). Since air quality directly affects human health, contributing significantly to the incidence of respiratory and cardiovascular diseases (Pope et al, 2002; Pope et al., 2009), straining health care facilities and driving health costs up, many megacities are directing intense efforts to improve air quality by controlling emissions.

In addition to local sources, many cities must also contend with pollution transported from surrounding areas. Upwind industrial activities can severely impact air quality. In the tropics, smoke from biomass burning may often exacerbate existing air quality problems in population centers [Pavagadhi et al., 2013; Marais et al; 2014; Yokelson et al., 2011; Kreidenweis et al., 2001]. At low latitudes high UV levels increase O₃ formation rates, and high temperatures result in high levels of Volatile Organic Compounds (VOCs) emissions and fast photochemistry. In addition to the issues of air quality within megacities, urban/industrial emissions from these cities can change the chemical content of the downwind troposphere, influencing both air quality and climate on scales ranging from regional to continental and global (Molina and Molina, 2002; Molina et al., 2004; Parrish et al., 2009; Molina et al., 2010).

In spite of the impact of pollution on air quality and human health there is no clear picture of the sources, sinks and distribution of major and minor pollutants in and around megacities. This is especially true of cities in the developing world, which have very limited monitoring capabilities. A few metropolitan areas do have surface measurement networks, but these cannot provide a “big picture” view of a region, or much information on the vertical distribution of pollutants. Furthermore, some chemical species have high spatial variability that a sparse network cannot capture. Aircraft campaigns, such as the Megacity Initiative: Local and Global Research Observations, MILAGRO (Singh et al., 2009), the Intercontinental Transport Experiment-Phase B, INTEX-B (Singh et al., 2009) in Mexico, and the African Monsoon Multidisciplinary Analysis (AMMA) (2006) yield valuable information on the vertical distribution of pollutants and their precursors, provide contextual background for air quality events, and significantly add to our understanding of regional processes; however, campaigns are expensive, and their temporal coverage is very limited. Satellite observations, with much greater temporal coverage than aircraft campaigns, and greater spatial coverage than in situ instruments, can fill the need for extensive data over long periods and large regions; these data can be used to determine the



larger scale spatial variability and the seasonal and interannual signals of atmospheric pollutants and provide useful constraints on model emissions and chemistry.

Here we will focus on data from the Tropospheric Emission Spectrometer (TES) flying on the NASA AURA satellite. TES data have been used extensively to evaluate and improve Chemical Transport Model (CTM) performance in previous air quality studies. TES ozone (O_3) data were assimilated into the Real-time Air Quality Modeling System (RAQMS) model and reduced model biases with respect to aircraft measurements during the Second Texas Air Quality Study (TexAQS II) (Pierce et al., 2009). Ammonia (NH_3) data from TES were used in an inverse modeling approach to improve emissions in the GEOS-Chem model, leading to better agreement with AMoN surface data (Zhu et al., 2013). Similarly, TES methanol (CH_3OH) data were also used to optimize GEOS-Chem methanol emission, leading to better agreement between the model and aircraft measurements (Wells et al., 2014), obtained from a number of campaigns, including MILAGRO (Singh et al., 2009), INTEX-B (Singh et al., 2009), the second Texas Air Quality Study, TexAQS-II (Parrish et al., 2009), and the Arctic Research of the Composition of the Troposphere from Aircraft and Satellites, ARCTAS (Jacob et al., 2010).

The studies above used measurements from two of TES observing modes that provide near-global coverage. To reduce wear and tear associated with the aging of the instrument and extend its lifetime, a new limited observing strategy was initiated in January 2013. This strategy reduces the number of sample points globally, but increases their density over designated regions, including 19 selected megacities (see Table 1). The TES megacity dataset contains simultaneous measurements (at approximately 1:30 pm local time) of a number of important trace gases, which provide a chemical snapshot of each city every two weeks, and can be used to determine the trace gas distribution and to evaluate model performance. In this study we use the megacity data to show variations of four key trace gases measured by TES, selected from the much larger set available: O_3 , NH_3 , CH_3OH and formic acid ($HCOOH$). Ozone is a secondary pollutant formed by the reaction of NO_x and VOCs, and is harmful to human and ecosystem health. CH_3OH is principally produced by plants; the main source of NH_3 is agricultural activity; $HCOOH$ is largely formed by photochemical oxidation of VOCs in the troposphere, but is directly emitted by vegetation and other sources (Millet et al., 2015). NH_3 , CH_3OH and $HCOOH$ are also by-products of biomass burning, and are correlated with CO during biomass burning events (Luo et al., 2015). While TES carbon monoxide (CO) has provided extremely useful information as a tracer for combustion in the past (Luo et al., 2007; Worden et al., 2013; Luo et al., 2015), current TES CO throughput is very poor (due to issues with the TES filter in the range sensitive to CO) and has not been included in this study.

We focus on two metropolitan areas, Mexico City, Mexico and Lagos, Nigeria. Both cities are located at low latitudes, but they lie in very different geographic contexts. Mexico City is landlocked at 2200 m altitude on the Tropic of Cancer, while Lagos lies at sea level on the Gulf of Guinea at six degrees north of the Equator. However, both are impacted to some degree by regular seasonal biomass burning in the



surrounding region. We use in situ measurements (in Mexico City) and back trajectories to assess the influence of biomass burning on the air quality within the cities themselves, and we demonstrate the use of the TES observations to provide finer scale information beyond that derived from chemical transport models in these regions.

- 5 Section 2 briefly describes the TES instrument and the characteristics of the TES megacity observations, as well as ancillary measurements and models used. Section 3 presents an analysis of the TES data over Mexico City and Lagos, focusing on how emissions from surrounding areas, including biomass burning and industrial sources, play a role in determining local air quality. Section 4 summarizes our results.

10 **2 Data and Models**

2.1 TES

TES is a nadir-viewing Fourier-transform infrared (FTIR) spectrometer with a high spectral resolution (0.06 cm^{-1}) and a nadir footprint of $5.3 \text{ km} \times 8.3 \text{ km}$ (Beer, 2006). TES flies aboard the NASA Aura spacecraft, which was launched in July 2004 and is part of the A-train constellation of satellites. It is in
15 a sun-synchronous orbit with an equator crossing time at around 01:30 and 13:30 local solar time. While TES does not provide the broad spatial coverage offered by scanning instruments such as the Metop Infrared Atmospheric Sounding Interferometer (IASI) or the NASA Atmospheric Infrared
20 Surveys (GS), its higher spectral resolution yields greater sensitivity to gas concentrations lower in the atmosphere (Clarisse et al., 2010). The standard mode of TES observations until 2011 was Global
25 Surveys (GS), which consisted of widely spaced observations ($\sim 180 \text{ km}$ apart) collected during 15 orbits (approximately 26 hours). During the gaps between GS observations, focused Special Observations were carried out, with various different objectives including collocation with aircraft campaigns (e.g., MILAGRO) and detailed studies of regions with significant pollution, such as eastern
30 China or the Fort McMurray area in Alberta. These SOs were more closely spaced (12 to 60 km), providing better information on regional variability. After 2011 SOs became the only operational mode. The TES Megacity observation strategy was implemented in January 2013 with the goal of providing data to improve our knowledge of the air quality over regions where a large fraction of the world's population lives. Each SO run, or transect, consists of 20 closely spaced footprints (12 km apart), running almost north to south (Figure 1), repeated every 16 days. Nineteen metropolitan areas (see
Table I) were chosen based on their population and known air quality problems. Table 1 contains the coordinates of the midpoint of each transect. Due to satellite viewing limitations, the transects do not necessarily cross directly over the center of each city.

TES retrievals follows an optimal estimation approach (Rodgers, 2000; Bowman et al., 2006) that minimizes the differences between the TES Level 1B spectra and a radiative transfer calculation that



uses absorption coefficients calculated with the AER line-by-line radiative transfer model LBLRTM (Clough et al., 2006). The optimal estimation algorithm produces the averaging kernels and error covariance matrices necessary for utilization of the data within data assimilation and inverse modelling frameworks. While the TES sensitivity to O₃ peaks between 600 and 400 mbar (Worden et al., 2007),
5 for the other trace gases discussed here the sensitivity peaks much lower in the atmosphere: between 900 and 700 mbar for NH₃ (Shephard et al., 2011) and CH₃OH (Cady-Pereira et al., 2012), and between 900 and 600 mbar for HCOOH (Cady-Pereira et al., 2014). The early afternoon overpass time is ideal for observing these gases, which tend to be mostly concentrated in the lower troposphere, as the thermal contrast between the surface temperature and the temperature at the peak of the TES sensitivity is
10 enhanced, providing stronger radiative signals.

In general results will be presented in terms of representative values instead of profiles: these values are calculated as the mean over the three pressure levels closest to the surface for O₃, and means of the three levels at and below the peak of the averaging kernel for NH₃, HCOOH and CH₃OH, which often coincide with the three near-surface levels. Only profiles with good quality flags are included in this
15 processing. The representative values are averaged over each transect to generate time series, and seasonally over each observation point to produce plots of spatial variability along each transect. The TES Megacity data are TES Version V006.

2.2 Other satellite data

20 Time series of CO V006 data from the NASA AIRS instrument, along with aerosol optical depth (AOD) from the NASA Moderate Resolution Imaging Spectroradiometer (MODIS) Deep Blue Algorithm are presented to provide context for the TES data. These are daily gridded products on a 1 deg x 1deg grid, obtained from the NASA Giovanni site (<http://giovanni.gsfc.nasa.gov/giovanni/>), and have been averaged over the length of the transect. Fire location maps were obtained from NASA Fire
25 Information for Resource Management System (FIRMS) site (<https://firms.modaps.eosdis.nasa.gov/firemap/>) using the MODIS NRT C5 product.

2.3 In-situ observations

In situ data from the Mexico City area are used to provide context for the satellite measurements. We
30 look at CO, as a tracer for combustion, and particulate matter (PM), both PM_{2.5}, which consists of particles with a diameter less than 2.5 microns, and PM₁₀, defined as particles with a diameter between 2.5 and 10 microns. These data were obtained from 28 stations (Figure 2) of the Air Quality Monitoring
Network of Mexico City (AQMN_MC) (Tzompa-Sosa et al., 2016)

35 (<http://www.aire.cdmx.gob.mx/default.php?opc=%27aKBhnmI=%27&opcion=Zg==>). Precipitation data were obtained from daily precipitation accumulation from 78 rain gauges across MCMA from the



Pluviometric Network of Mexico City (<http://www.sacmex.cdmx.gob.mx/sacmex/>). We will also refer to an analysis carried out by Tzompa-Sosa et al. (2016), which used measurements of levoglucosan (LEV) and water soluble organic carbon (WSOC) from these same sites. The mass ratio of LEV to WSOC determined from source samples can be used to determine the contribution of primary biomass smoke to the total WSOC concentrations in ambient samples (Sullivan et al., 2008).

2.5 Model output

We used model output to provide background information and context for the TES measurements. HYSPLIT runs provided back trajectories, the MIROC model (see section 2.5.2) yielded a regional view of O₃ and CH₃OH in the MCMA area, and the nested GEOS-Chem runs provided O₃, NH₃, CH₃OH and HCOOH at higher spatial resolution in the Lagos area.

2.5.1 HYSPLIT runs

To explore the sources influencing the trace gas mixing ratios measured by TES over Mexico City and Lagos we generated back trajectories using Version 4 of Hybrid Single-Particle Lagrangian Integrated Trajectory (HYSPLIT) model (Draxler and Hess, 1998). We used 1 deg x 1 deg gridded meteorological inputs from the Global Data Assimilation System (GDAS) to generate 5-day back trajectories from Mexico City and 10-day back trajectories from Lagos. The longer back trajectory length over Lagos was used to account for the slower wind speeds near the equator. The back trajectories at each location were initiated from 100, 500, 1000, 2000, 3000, and 4000 m above ground level to identify the potentially different source locations impacting the vertical profiles of pollutants in the lower troposphere.

2.5.2 MIROC model output

The MIROC-Chem model (Watanabe et al., 2011) estimates detailed photochemistry in the troposphere and stratosphere by simulating tracer transport, wet and dry deposition, and emissions, and calculates the concentrations of 92 chemical species and 262 chemical reactions. Its tropospheric chemistry was developed based on the CHASER model (Sudo et al., 2002) and it resolves the fundamental chemical cycle of Ox-NO_x-HO_x-CH₄-CO along with oxidation of NMVOCs to simulate ozone chemistry in the troposphere. However NH₃ emissions are limited and HCOOH is not modeled separately. Its stratospheric chemistry was developed based on the CCSR/NIES stratospheric chemistry model (Akiyoshi et al., 2004). MIROC-Chem has a T42 horizontal resolution (approximately 2.8°x2.8°) with 32 vertical levels from the surface to 4.4 hPa. It is coupled to the atmospheric general circulation model MIROC-AGCM version 4 (Watanabe et al., 2011). The MIROC-AGCM fields were nudged toward the 6-hourly ERA-Interim reanalysis (Dee et al., 2011). The emission inventories used were obtained from



the emission scenarios for Greenhouse Gas and Air Pollution Interactions and Synergies (GAINS) model developed by International Institute for Applied System Analysis (IIASA) (Klimont et al., 2009; Akimoto et al., 2015).

In this work the model employed daily surface NO_x and CO emission data optimized using satellite
5 measurements (OMI, GOME-2, and SCIAMACHY NO₂ measurements for NO_x emissions, and MOPITT CO measurements for CO emissions) for May 2013 from a data assimilation calculation (Miyazaki et al., 2016). The optimized emissions are about three times larger over Mexico city and about 20 % larger for the United States total emissions in May 2013 than the a priori emissions constructed based on bottom-up emission inventories from the Emission Database for Global
10 Atmospheric Research (EDGAR) version 4.2 (EC JRC/PBL, 2012), the Global Fire Emissions Database (GFED) version 3.1 (van der Werf et al., 2010), and the Global Emissions Inventory Activity (GEIA) (Yienger and Levy, 1995). These results suggest that the bottom-up emission inventories underestimate NO_x emissions and may lead to a large uncertainty in ozone simulations around Mexico City.

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2.5.3 GEOS-Chem output

The GEOS-Chem chemical transport model (version 10-01; <http://acmg.seas.harvard.edu/geos/>) uses as input GEOS-5 assimilated meteorology from the NASA Global Modeling and Assimilation Office (GMAO) at $0.5^{\circ} \times 0.667^{\circ}$ horizontal resolution nested over the African continent. Boundary conditions
20 are from a global simulation at $2^{\circ} \times 2.5^{\circ}$. The model includes detailed NO_x-VOC-ozone-aerosol photochemistry described in Mao et al. (2010) and updated here to include sea-air exchange of methanol as in Wells et al. (2014). Pollution sources relevant to Lagos, Nigeria include seasonal biomass burning from GFED4 (van der Werf et al., 2010), industrial emissions from EDGAR v4.2 (EC-JRC/PBL, 2011) for NO_x and CO and RETROv2 for VOCs (Schultz et al, 2007), NH₃ from agricultural
25 activity from EDGAR v4.2, pollution from residential biofuel (fuelwood, charcoal, crop residue) use, charcoal production, backup generators, cars, and motorcycles if from the regional DICE-Africa inventory (Marais and Wiedinmyer, 2016), and trash burning from Wiedinmyer et al. (2014). Data are for year 2012 with one year of spinup for chemical initialization.

3 TES Megacity data examples

30 3.1 Mexico City

The Mexico City Metropolitan Area (MCMA) is home to over 20 million people and is situated in a basin 2240 m above sea level, surrounded by mountains on three sides. To the north lies the Mexican Plateau, while there is a small gap in the southeastern rim. At approximately 19.4°N, it is at the edge of the tropics and receives strong solar radiation that drives mountain-valley and urban-induced winds (de



Foy et al., 2008), leading to convergent drainage flow and associated high concentrations of pollutants (Jauregui, 1988). Aircraft campaigns have linked Mexico City pollution to biomass burning emissions originating outside the metropolitan area (Yokelson et al., 2011) and possibly even from the Yucatan peninsula (Kreidenweis et al., 2001). Here we use TES data to look for connections between biomass
5 burning and pollution in the MCMA (Figure 3) during the 2013-2015 period. The time series in the left panel of Figure 3 shows a distinct seasonal variability in all four species we are analyzing, with enhanced values between March and June of each year. This period coincides with the dry season and concomitant biomass burning events in Mexico, as shown by the time series of AIRS CO, a marker for combustion, and MODIS AOD; high aerosol AOD may indicate the presence of smoke and/or dust
10 (right panel of Figure 3). The seasonal cycle in the latter two datasets is qualitatively in good agreement with the cycle presented in all four TES time series.

The strongest peak in the time series for all species occurs on May 9, 2013 (indicated by red dashed line in Figure 3). This extreme event detected by TES occurs within a period of several days of maximum values in the AIRS and MODIS record for the 2013-2015 period. The concurrent peaks in CO, AOD,
15 and HCOOH, which is a known product of biomass burning, all strongly indicate biomass burning influence in the MCMA on this day, as does the very high of HCOOH (nearly 5 ppbv). We compared the wind direction and fire locations for May 9 with those on May 25, 2013 (green dashed line in Figure 3), the next TES MCMA transect, when abundances of all of the trace gases were much lower (Figure 4). Back trajectories were calculated with the NOAA HYSPLIT model, while the fire locations were
20 obtained from the NASA FIRMS site. The back trajectories for May 9 show that the air masses arriving on that day in MCMA passed over or near a number of fires west and southwest of MCMA. Conversely, the trajectories on May 25 mainly traversed the Gulf of Mexico over a region with few detectable fires before climbing to the MCMA plateau. The combined satellite data and back trajectories suggest that the air quality along the transect on May 9 was likely impacted by biomass
25 burning, but on May 25 these same data, in particular the ~3 ppbv lower value of HCOOH, indicate that transport from biomass burning emissions probably had much less of an impact, with local production possibly being responsible for the lower but still above average abundances detected on this day. The TES measurement on April 23 (dashed blue line in Figure 3) is interesting in that it contains elevated O₃, CH₃OH and HCOOH, all of which could have a fire source, but low NH₃. Fire maps show a
30 number of fires around Mexico City on this day; however, the low values of NH₃, which has a shorter life than the other species, could indicate that these fires were further away than on May 9. By June 10 (brown dashed line in Figure 3) the TES transect (the second point after the May 9 peak), shows low values for all species, indicating cleaner air.

Surface measurements of precipitation, PM_{2.5}, PM₁₀ and CO (Figure 5) show that the late April-early
35 June period in 2013 can be divided into two regimes: a dry regime from April 27 through May 11, with higher concentrations of PM_{2.5}, PM₁₀ and CO, peaking around May 9, followed by a wet regime



through June 2, with lower pollutant concentrations and a number of rainy days. Tzompa-Sosa et al. (2016) calculated LEV/WSOC ratios during May 2013, which provide an estimate of the contribution of biomass burning to the total WSOC in an air mass. Their results show higher LEV/WSOC ratios during the first weeks of May, compared to the rest of the month. Taken altogether, the data indicate

5 that biomass burning contributes much more to pollution on May 9 than May 25, due to a combination of transport and the lack of precipitation to drive wet deposition. It is, in fact, likely that the changing meteorological conditions associated with the wetter weather were also responsible for the shift in transport away from the fire locations. Note that on June 10th, when TES data suggest cleaner air, the in situ measurements show rain and lower values of CO and particulate matter.

10 However, it is important to note the impact of biomass burning varies spatially. Figure 6 shows that NH₃ and HCOOH concentrations are extremely high to the south of the Mexico City basin on May 9 and drop off markedly as TES moves north of 19.1° and elevation increases; on May 25 the decrease in NH₃ and HCOOH with elevation is much more gradual, and there is actually an increase as the TES observes the air mass inside the basin. The overall abundances of NH₃ and HCOOH are higher on May

15 9 than May 25, which is consistent with the in situ observations, but the difference between the days is much more marked south of the caldera. These results suggest that on May 9 the air inside the basin was somewhat isolated from the biomass burning influence due to the fact that the air from the fires had to travel from the southwest up and over the mountains to reach MCMA (Figure 4). Seasonal means of NH₃ and HCOOH (Figure 7) along the transect path show that this a regular feature, especially in the

20 December through May period: concentrations decrease sharply as the transect passes north over the highest elevation point, at 3088 m, just south of Mexico City, though the MAM HCOOH mean reaches its minimum slightly further north.

Both O₃ and CH₃OH are generally higher on May 9 than on May 25, confirming the high pollution level of this day as indicated by NH₃ and HCOOH. On the other hand, the spatial variability along the

25 transects for O₃ and CH₃OH is very different than for the other species. O₃ is relatively flat, and CH₃OH on May 9 has higher values and variability within the caldera rather than to the south. It is not clear if the weak spatial variability in O₃ is due to a possible lack of TES sensitivity to near surface amounts or has a physical origin, nor are the drivers of the CH₃OH variability at all obvious. These questions are left for future work.

30 We examined the capability of the MIROC model to capture the different conditions on these two days. Due to its coarse resolution MIROC cannot model the spatial variability seen in the TES data, so we compared the average over the transects against the model grid cell containing Mexico City. On May 9 (Figure 8, top panels) TES data show that above 800 mbar the model severely underestimates both species on this day. These underestimations could be attributed to transport errors associated with the

35 relatively coarse model resolution (i.e., 2.8°) and to errors in photochemical productions for such an extremely polluted case. On May 25 (Figure 8, bottom panels), when biomass burning appears to play a



much smaller role, there is better agreement between TES and the MIROC profiles, though either the model has overestimated CH₃OH below 800 mbar or TES was not sufficiently sensitive to CH₃OH at this altitude range. O₃ is slightly overestimated below 600 mbar. The overall good model performance is largely attributable to the use of optimized anthropogenic and biomass burning emissions of NO_x and CO using satellite measurements (Miyazaki et al., 2016), but this example demonstrates the value of the TES data in capturing events at finer scales than large scales are capable of estimating..

3.2 Lagos

The atmosphere above urban West Africa is very poorly understood, in spite of the large and growing population. Not only do the pollution levels impact the health of tens of millions of people, they very likely alter the precipitation patterns of the West African Monsoon (Knippertz et al. 2015). While there have been a number of measurement campaigns in the region, such as the African Monsoon Multidisciplinary Analysis (AMMA) (Mari et al., 2011), the IGAC (International Global Atmospheric Chemistry)/DEBITS (Deposition of Biogeochemically Important Trace Species)/AFRICA atmospheric chemistry and deposition monitoring network IDAF (<http://idaf.sedoo.fr>; in operation since 1995, Liousse et al., 2010) and the Aerosol Robotic Network (AERONET), these have focused on natural emissions in remote locations. For example, the closest AERONET station to Lagos is at Ilorin, almost 300 km away in the sub-Sahel. There is very little in situ monitoring within the Lagos region. This unavailability of local data increases the value of the TES data, as it provides consistent (i.e., from the same platform at the same time) and repeated (every two weeks) measurements of a number of species over a 200 km extent. The data can point to interesting processes and trends, and provide context for the few in situ measurements available, and their coarser spatial resolution enables better comparisons to models than sparse in situ observations (Schutgens et al., 2016).

Lagos is a tropical city at 6.5° N on the northern edge of the Gulf of Guinea. With over 12 million people, Lagos is Nigeria's and Africa's largest city and one of the fastest growing megacities, with a growth rate currently at 4%/year (UN, 2014). This rapid growth has had many consequences, including traffic congestion, high use of very inefficient motorcycles with two-stroke engines, unreliable electrical supply, and poor waste disposal, which in turn have led to high emissions from vehicles, diesel powered generators and waste incineration (Hopkins et al., 2009; Marais et al., 2014). The oil and gas extraction activities in the Niger Delta to the southeast also contribute to poor air quality through flaring, gas leakage and pipeline explosions (Marais et al., 2014). At this low latitude synoptic variability is limited. Air quality in the Lagos region is strongly affected by the West African Monsoon (WAM) during the wet season (June-August (JJA)), which effectively ventilates the area (Marais et al., 2014) and reduces the aerosol load through precipitation (Knippertz et al., 2015). In the dry season (December-February), strong temperature inversions at 900-750 hPa inland, caused by the warm northeasterly Harmattan



winds and the blocking anticyclone from the Sahara (Sauvage et al., 2005), trap pollutants in the lowest levels of the atmosphere. The ventilation levels during this season are among the weakest in the world (Marais, 2014). There is frequent recirculation of air due to sea breezes, an example of which is shown in the back trajectory maps for February 7, 2013 (Figure 9). Anthropogenic pollutants can build up and interact with the products from frequent biomass burning events (Figure 9). Sunlight driven chemistry can produce ozone and acids (H_2SO_4 , HNO_3); the latter react with ammonia to create $\text{PM}_{2.5}$ particles which contribute to the aerosol loading.

TES transects were carried out slightly to the west of the Lagos area (Figure 10); the southern end of the transect, south of 6.2° , was over the Gulf of Guinea. The elevation is low and constant; there is no complicated topography affecting winds and transport as there is in the MCMA. Means of the TES observations along the Lagos transect (Figure 11, left panel, solid line) show a strong seasonal pattern in all four species, with clear enhancements during the December-March period, which correlate with the AIRS CO and MODIS AOD seasonal variability (Figure 11, right panel). Comparing the Lagos time series to those from the the MCMA (Figure 11, left panel, dashed line), it becomes obvious that the air quality problems of these two regions are different in magnitude. Mexico City has a reputation for severe pollution, yet the high levels of all four trace gases measured during the extreme event on May 9 in the MCMA are actually below the regular dry season levels in Lagos, reflecting the far greater number and strength of pollution sources, from biomass burning, traffic, power generation and the petrochemical industry, combined with the stagnation and recirculation described above. This demonstrates the potential of the TES megacity data to quickly evaluate the similarities and differences in the air quality components of the world's megacities.

Further evidence of the recirculation of trace gases is provided in the spatial variability of the seasonal means over the 2013-2015 period (Figure 12, top row). During the DJF period the values over the Gulf are approximately 90 ppbv, 4 ppbv, 4 ppbv and 4.5 ppbv for O_3 , NH_3 , CH_3OH and HCOOH , respectively; these values are a great deal higher than typical remote ocean values (i.e., 30 ppbv, 0.1 ppbv, 1.0 ppbv, 0.01 ppbv from GEOS-Chem global runs) for the same species, and are in general as high as (except for NH_3) the amounts over land, even though there are no significant sources of any of these species in the Gulf. NH_3 does show a sharp increase as the transect crosses from ocean to land, perhaps because some of the NH_3 carried out to sea has reacted with the nitric and sulphuric acids derived from NO_x and SO_2 emissions. NH_3 also exhibits this spatial gradient in the other three seasons, especially the MAM period. CH_3OH and HCOOH show steep decreases in JJA and SON as the transects moves north of 7°N , into less urbanized areas, indicating a strong city source for these gases in the absence of biomass burning. There is also a dip in the O_3 concentrations in DJF and MAM between 6.4 and 7°N , as the transect approaches the coast and passes along the Lagos urban area, possibly due to higher NO_x trapped in the boundary layer. The seasonality in the TES measurements is also evident in



these figures; the highest values are observed in DJF and MAM, with little difference between the two periods for CH_3OH and HCOOH , and larger differences for O_3 and NH_3 . Measurements are much lower in JJA and SON seasons.

Comparisons with model output are per force qualitative, as the fire emissions and meteorological conditions between the GEOS-Chem 2012 runs and the TES 2013-2015 measurements are likely quantitatively different; the seasonal patterns would be expected to show similarities, but the TES spatial variability should be higher, as the TES pixel is twenty five times smaller than the GEOS-Chem nested grid box. The GEOS-Chem spatial plots (Figure 12, bottom row) were obtained by averaging the GEOS-Chem mean seasonal profile values over the same vertical range used to provide the TES spatial plots. This vertical range varied slightly by season and latitude, due to variability in the TES sensitivity for all species except O_3 , for which the range was fixed to cover the lowest three measurement levels. We used the TES vertical averaging interval in order to reduce differences arising from averages over different numbers of levels. TES and GEOS-Chem both see high values in DJF, lower values in JJA and SON, and a latitudinal gradient in NH_3 in DJF, although the GEOS-Chem gradient is much weaker. GEOS-Chem predicts ocean-land gradients for all species and seasons, but these are only observed in the TES NH_3 measurements, suggesting that the model may be underestimating transport of pollution off-shore, particularly during DJF and MAM, either because of differences in meteorology between the modelled and measured time periods or because of deficiencies in modelling of the local scale land-sea circulation. GEOS-Chem values are lower by a factor of 2 (for O_3) to 10 (for NH_3), except for CH_3OH over land in DJF, where both model and measurement predict ~ 4 to 5 ppbv. Chaliyakunnel et al. (2016) have shown that there are secondary sources of HCOOH from fires that are not well modelled, which could be contributing to the underestimate of HCOOH . However, since these large differences are evident in the HCOOH data in the non-burning season, and are also seen in O_3 , and NH_3 , the city of Lagos itself is probably a strong pollution source, due to the conditions described at the beginning of this section.

4 Summary

The TES Megacity dataset covers the period from January 2013 to present and provides air quality data over nineteen metropolitan areas. We focused on O_3 , NH_3 , CH_3OH , and HCOOH for their role in determining air quality.

In the Mexico City region the TES-observed seasonal signal is qualitatively well correlated with AIRS CO and MODIS AOD, with peaks in the drier MAM period, which is also the peak of the biomass burning season. A very high pollution event on May 9, 2013, was observed and attributed to contributions from biomass burning emissions transported from the southwest. This event was observed in Mexico City itself, where high LEV/WSOC ratios were measured, but the TES data showed it was



significantly stronger south of the city. This pattern of higher levels of these trace gases south of the caldera was repeated in MAM over the 2013–2015 period monitored by TES. A weaker event on May 25, 2013, in a rainier period with easterly winds, showed lower biomass burning influence both in the TES data, which exhibited lower HCOOH levels than on May 25, and in the in situ LEV/WSOC ratios, which were also significantly lower.

The TES data over Lagos showed a strong seasonal signal correlated with AIRS CO and MODIS AOD, peaking in the dry DJF period, when biomass burning events are most frequent and western Africa often experiences stagnant air conditions. Except for NH₃, TES did not observe gradients between the Gulf and the coast, even though these were predicted by the GEOS-Chem model. As sources of all these species are weak over the ocean, the abundances observed by TES are likely due to stagnant air masses being driven by sea breezes back and forth between the Gulf and the continent, suggesting that GEOS-Chem is not modeling the local transport correctly. The levels of trace gases measured by TES were noticeably higher over Lagos than over Mexico City, and these higher levels were more persistent. This points to a difference in air quality between the two cities: while Mexico City has pollution events in MAM, Lagos has almost continuous high levels of pollution in DJFM.

In summary, the TES Megacity dataset can provide seasonal variability and spatial gradients of many species relevant to air quality for cities around the world. The data can be used for process studies and model evaluation, and allows for easy comparisons between two or more cities.

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Table 1: TES Megacity Transect Midpoint Coordinates

City	Lat	Lon
Bangkok	13.6383	100.304
Beijing	39.8543	116.386
Buenos Aires	-34.7112	-58.9112
Delhi	28.957	77.4496
Dhaka	23.6374	90.1974
Houston	29.7203	-95.2691
Istanbul	40.9605	29.1336
Karachi	24.6877	66.7348
Kolkata	22.5168	88.4081
Lagos	6.57795	3.25456
Los Angeles	34.0724	-118.146
MexicoCity	19.1627	-99.2384
Mumbai	18.8821	72.8437
New York City	40.7045	-73.9673
Paris	48.8499	2.37268
SWChina	36.9589	114.249
Sao Paulo	-23.5372	-46.6846
Shenzhen	22.3653	113.674
Tokyo	35.5149	139.425



Figures

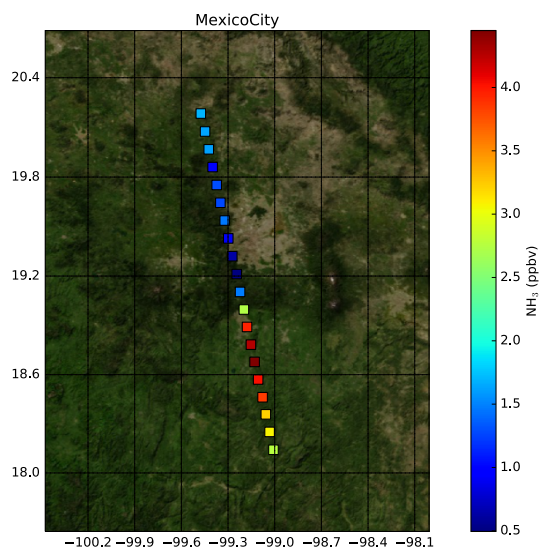


Figure 1: Sample TES NH_3 transect over the MCMA region.

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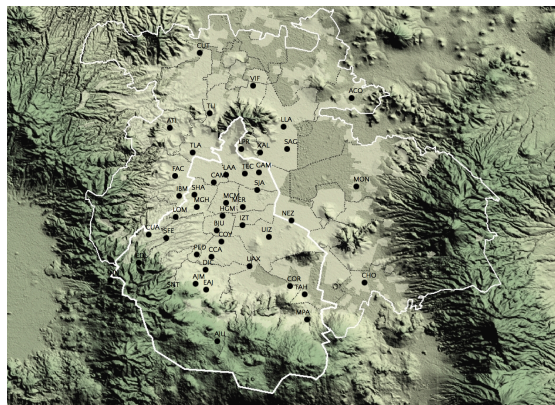
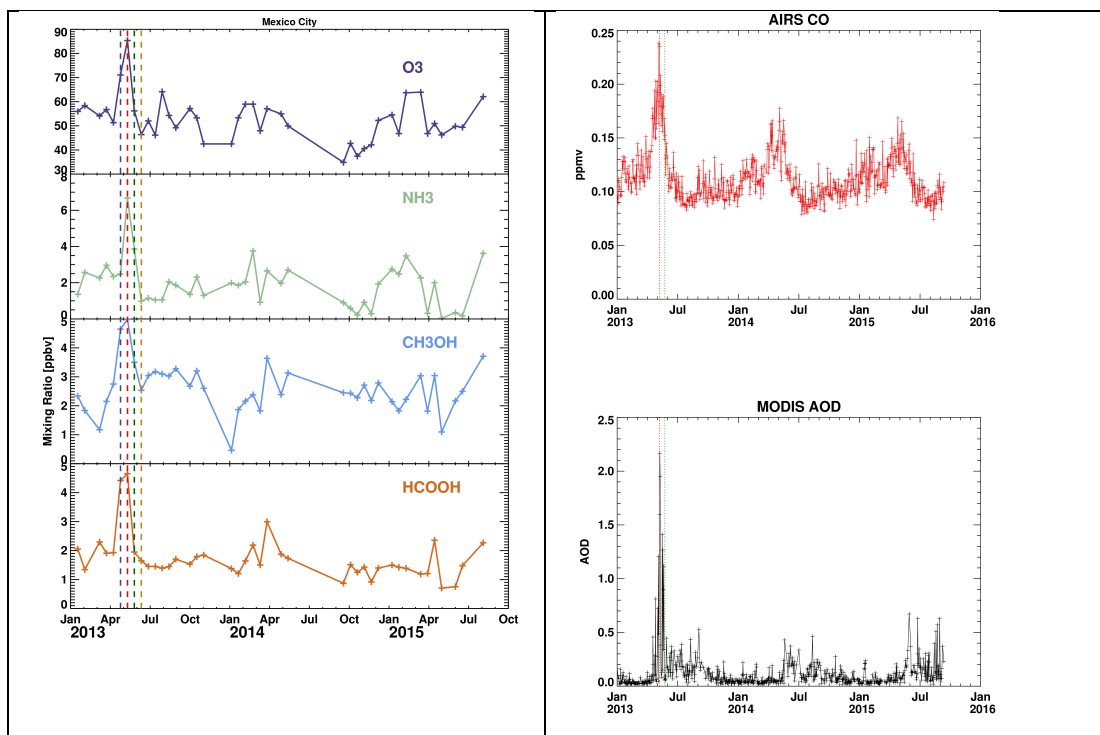
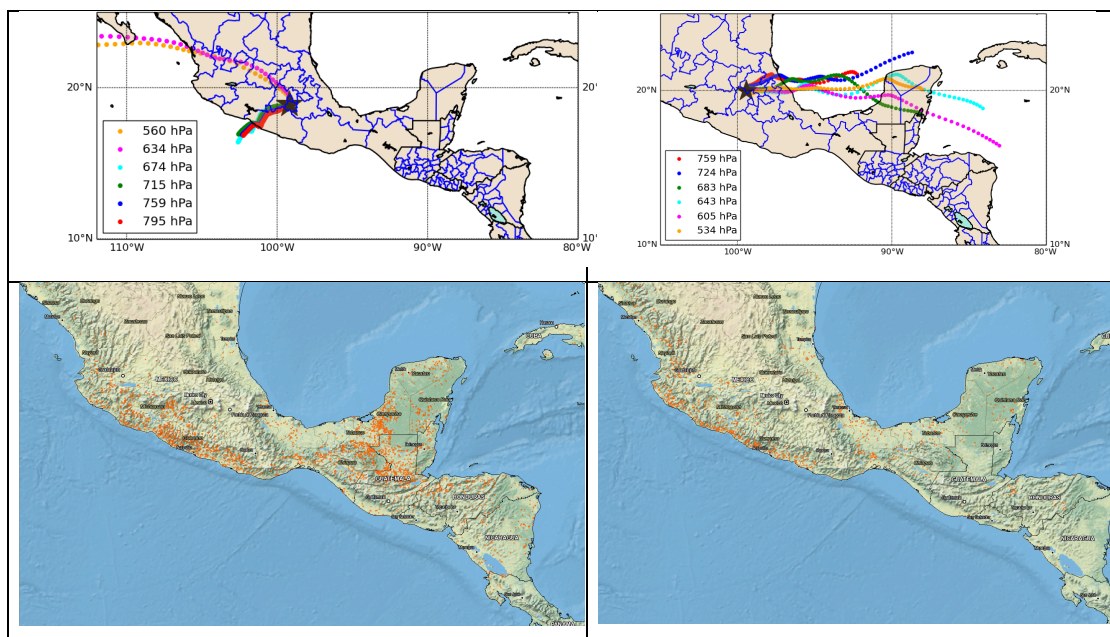


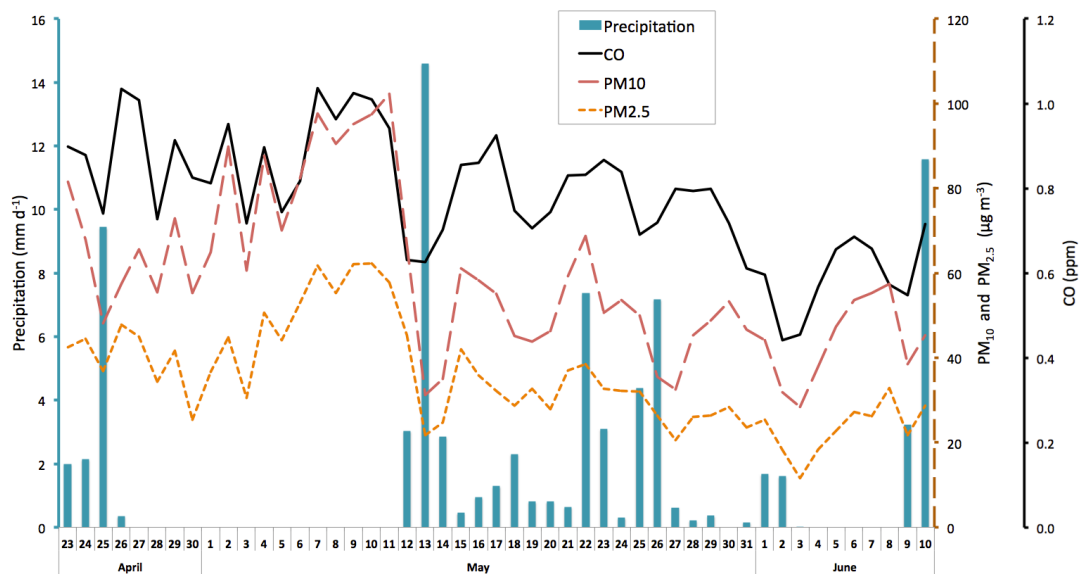
Figure 2: Locations of monitoring sites in the MCMA area.



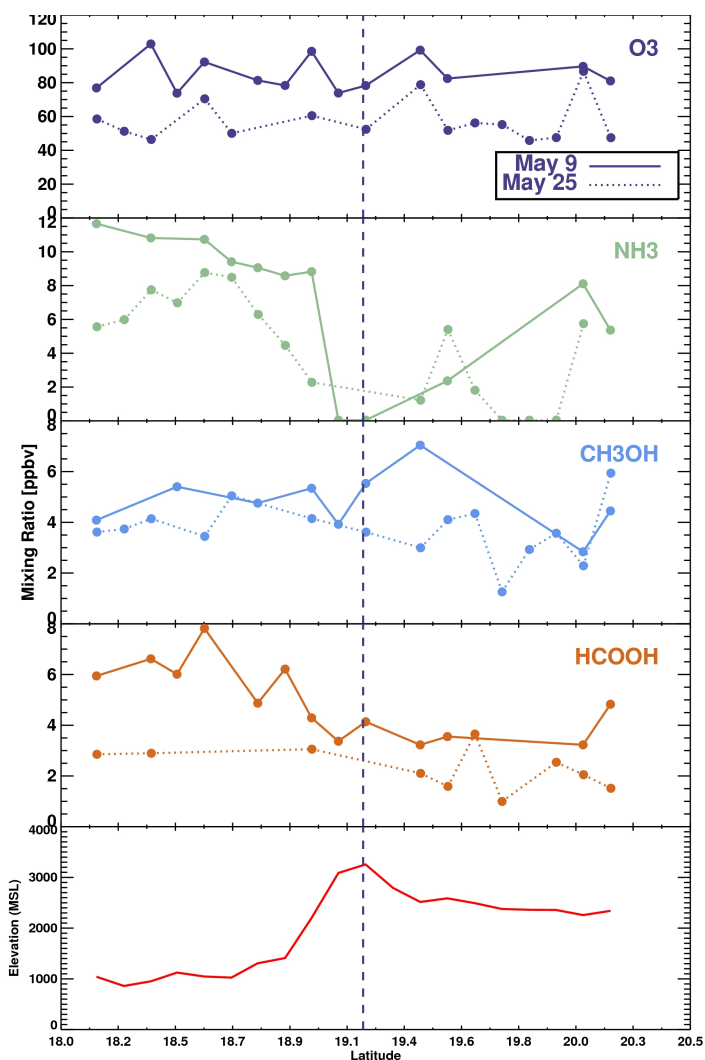
5 Figure 3: (left) Means of the lower levels (see text for definition) of each TES transect over the MCMA area; dashed lines mark dates discussed in text (right) AIRS CO @850 hPa (top) and MODIS Dark Target AOD (bottom) 1 deg x 1 deg data averaged over the grid boxes traversed by the transect.



5 **Figure 4: HYSPLIT back trajectories (top) and FIRMS fire locations (bottom) in Mexico for May 9 (left) and May 25 (right). Fire maps include fires from the selected dates and the two previous days. Pressures on back trajectory plots refer to level at which the back trajectory was initiated.**



10 **Figure 5: Averaged daily precipitation and ambient concentrations of $PM_{2.5}$, PM_{10} , and CO over the MCMA. Averaged daily precipitation from 78 stations located inside the MCMA. Averaged ambient concentrations of $PM_{2.5}$, PM_{10} , and CO over the MCMA correspond to 28 stations located inside the MCMA (see Figure 2).**



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Figure 6: TES retrievals versus latitude on May 9 (solid) and May 25 (dotted) 2013.: O₃ (top panel), NH₃(second panel), CH₃OH (third panel), HCOOH (fourth panel); elevation (bottom panel). Dashed line indicates latitude of the edge of the caldera containing Mexico City and elevation is maximum.

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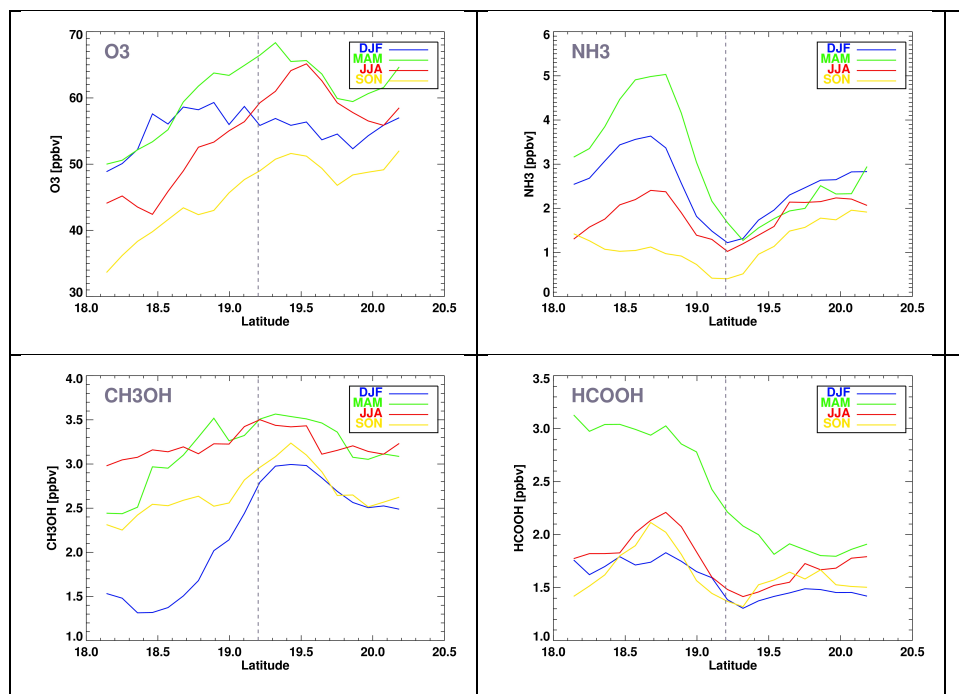


Figure 7: Seasonal means along the MCMA transect: O₃ (top left), NH₃ (top right), CH₃OH (lower left), HCOOH (lower right).

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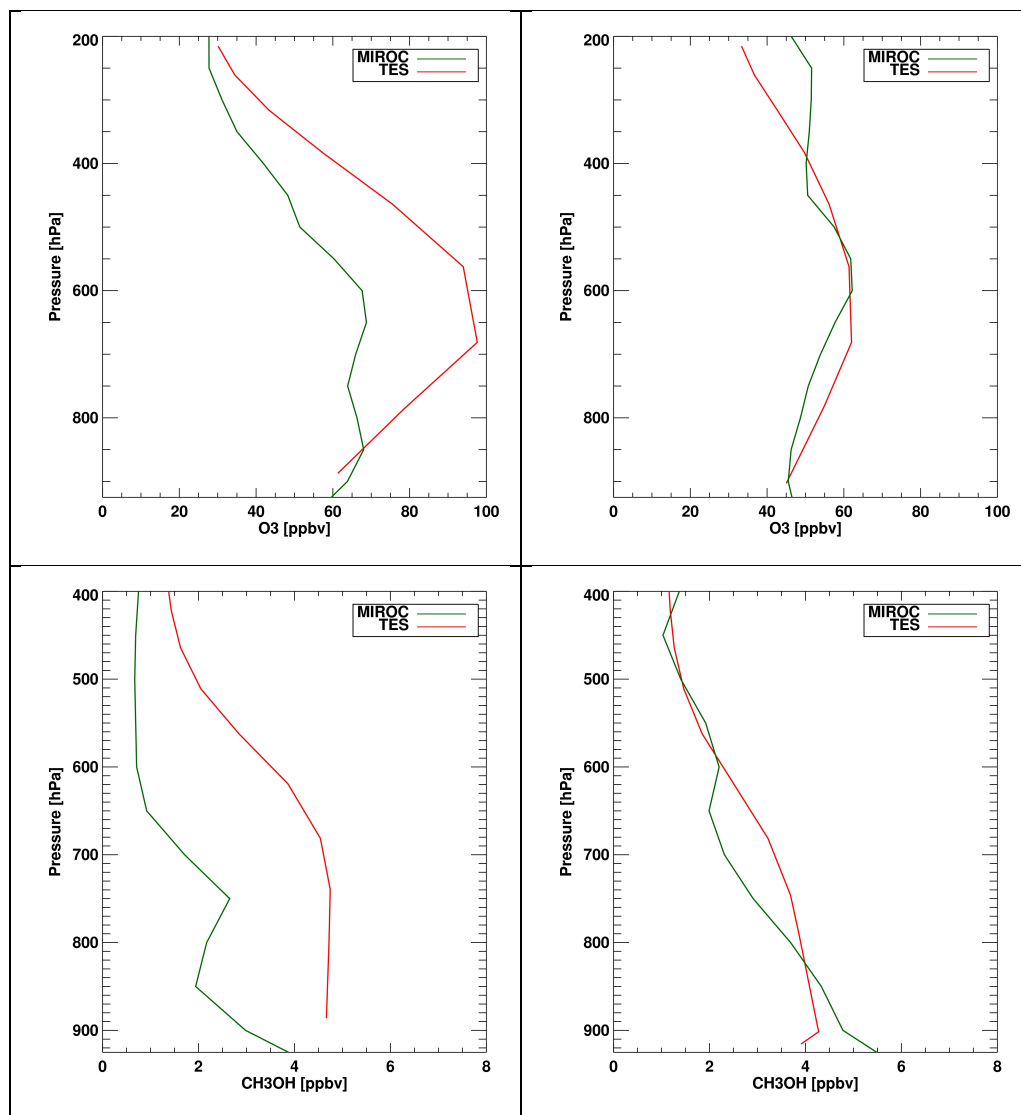


Figure 8: MIROC model and TES retrieved profiles for O₃ (top) and CH₃OH (bottom) on May 9 (left) and May 25 (right).

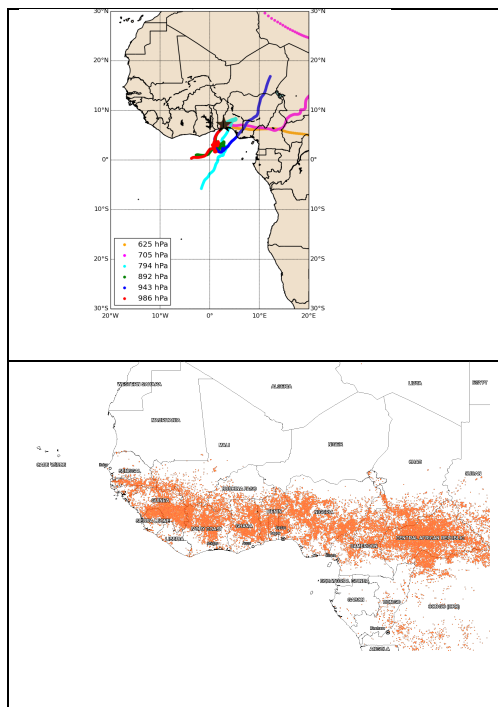


Figure 9: HYSPLIT back trajectories (top) and FIRMS fire locations (bottom) over western Africa for February 7, 2013. Fire map include fires from February 7 and the previous ten days.

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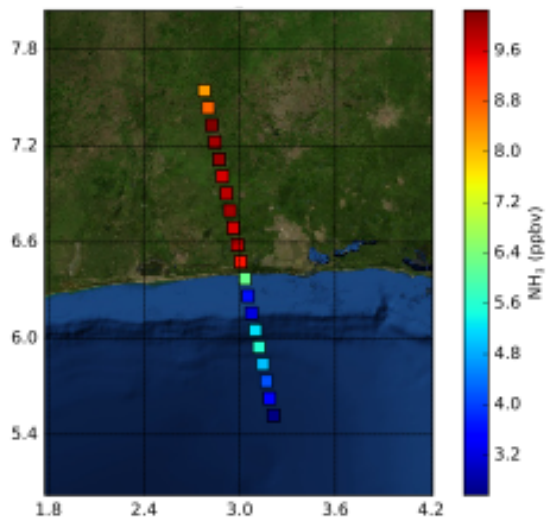


Figure 10: Sample TES NH_3 transect in the Lagos region in DJF; Lagos is visible as the brown area east of the transect near the coast.

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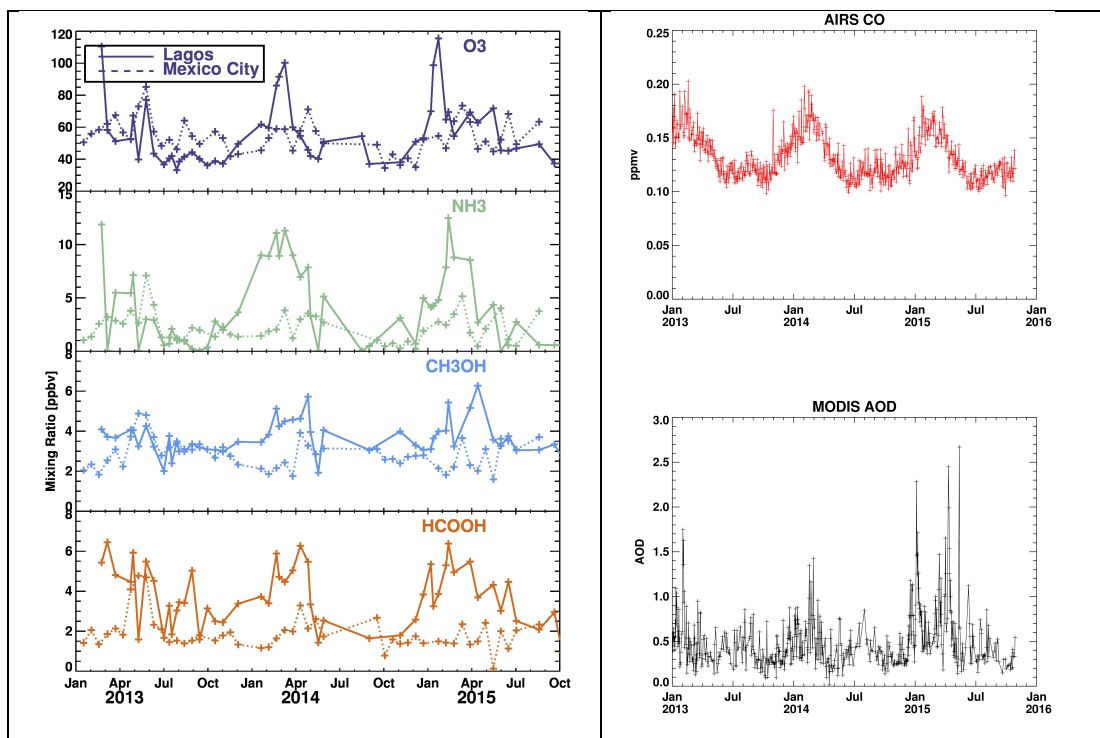


Figure 11: Means of the lower levels of each transect over the Lagos transects area (solid) and MCMA (dashed); (right) AIRS CO @1000 hPa (top) and MODIS Dark Target AOD (bottom) 1 deg x 1 deg data averaged over the grid boxes crossed by the transect.

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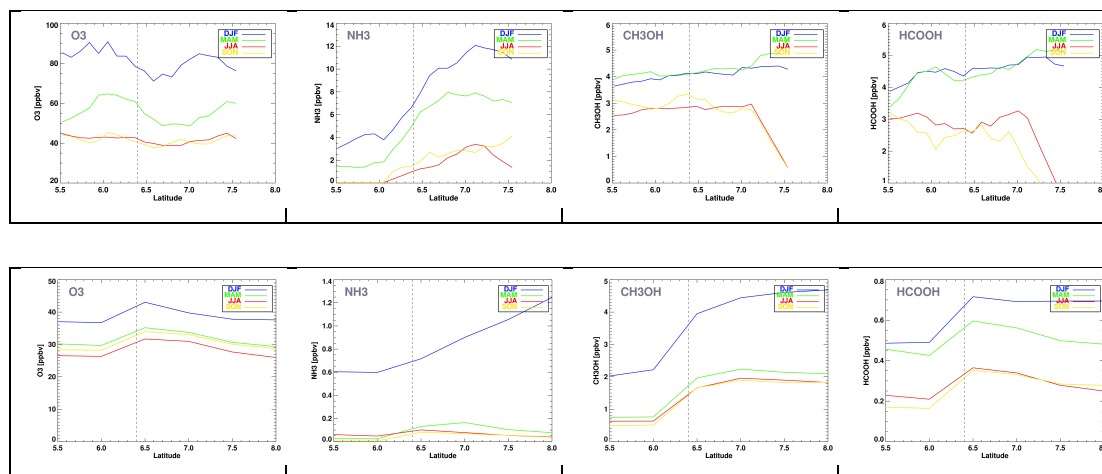


Figure 12: Seasonal means along the Lagos transect. TES 2013-2015 means (top), GEOS-Chem 2012 means (bottom). Note different y-axis vertical range between TES and GEOS-Chem plots.

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