Vehicle fleet emissions of black carbon, polycyclic aromatic hydrocarbons, and other pollutants measured by a mobile laboratory in Mexico City


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Abstract. Black carbon (BC) and polycyclic aromatic hydrocarbons (PAHs) are of concern due to their effects on climate and health. The main goal of this research is to provide the first estimate of emissions of BC and particle-phase PAHs (PPAHs) from motor vehicles in Mexico City. The emissions of other pollutants including carbon monoxide (CO), oxides of nitrogen (NOx), volatile organic compounds (VOCs), and particulate matter of diameter 2.5 µm and less (PM2.5) are also estimated. As a part of the Mexico City Metropolitan Area field campaign in April 2003 (MCMA-2003), a mobile laboratory was driven throughout the city. The laboratory was equipped with a comprehensive suite of gas and particle analyzers, including an aethalometer that measured BC and a photoionization aerosol sensor that measured PPAHs. While driving through traffic, the mobile lab continuously sampled exhaust plumes from the vehicles around it. We have developed a method of automatically identifying exhaust plumes, which are then used as the basis for calculation of fleet-average emissions. In the approximately 75 h of on-road sampling during the field campaign, we have identified ~30 000 exhaust measurement points that represent a variety of vehicle types and driving conditions. The large sample provides a basis for estimating fleet-average emission factors and thus the emission inventory. Motor vehicles in the Mexico City area are estimated to emit 1700±200 metric tons BC, 57±6 tons PPAHs, 1 190 000±40 000 tons CO, 120 000±300 tons NOx, 240 000±50 000 tons VOCs, and 4400±400 tons PM2.5 per year, not including cold start emissions. The estimates for CO, NOx, and PPAHs may be low by up to 10% due to the slower response time of analyzers used to measure these species. Compared to the government’s official motor vehicle emission inventory for the year 2002, the estimates for CO, NOx, VOCs, and PM2.5 are 38% lower, 23% lower, 27% higher, and 25% higher, respectively. The distributions of emission factors of BC, PPAHs, and PM2.5 are highly skewed, i.e. asymmetric, while those for benzene, measured as a surrogate for total VOCs, and NOx are less skewed. As a result, the total emissions of BC, PPAHs, and PM2.5 could be reduced by approximately 50% if the highest 20% of data points were removed, but “super polluters” are less influential on overall NOx and VOC emissions.

1 Introduction

Mexico City has become known for its air pollution problem as a result of the rapid growth of population, industry, and services, which encouraged an enormous increase in transportation activity and related pollutant emissions. In the 1990s, the Mexican government implemented pollution control measures on vehicles and fuels, which successfully reduced the ambient concentrations of three criteria pollutants: lead (Pb), carbon monoxide (CO), and sulfur dioxide (SO2). Nonetheless, air quality standards for other pollutants are still frequently violated. For example, the 1-h ozone standard of 110 ppb was exceeded on ~80–90% of the days every year between 1988 and 2000, and the daily PM10 (particulate
matter of aerodynamic diameter 10 $\mu$m and less) standard of 150 $\mu$g m$^{-3}$ was violated on more than 20% of the days in 1995–1998 (Molina and Molina, 2002). As in most large cities, the transportation sector in Mexico City is a major source of air pollution. The vehicle population is estimated to be 3.6 million (Comisión Ambiental Metropolitana, 2004), and factors such as congestion, lack of emission controls on many vehicles, and poor fuel quality contribute to higher vehicle emissions (Gakenheimer et al., 2002). According to the government’s 2002 emission inventory, mobile sources contribute over 99% of all CO, 84% of nitrogen oxides (NO$_x$), 39% of hydrocarbons (HC), 58% of SO$_x$, 19% of PM$_{10}$, and 52% of PM$_{2.5}$ emitted in the Mexico City Metropolitan Area (MCMA) (Comisión Ambiental Metropolitana, 2004). A five-week field campaign was conducted in April 2003 to support the understanding of the air pollution problem in Mexico City. In this work, we focus on motor vehicle emissions of black carbon (BC) and particulate polycyclic aromatic hydrocarbons (PPAHs), whose emissions have not yet been systematically estimated.

Black carbon refers to the elemental carbonaceous component of particulate matter that is formed through incomplete combustion of organic substances. Besides its direct effect on visibility, BC also influences global and regional climate and public health. Recent modeling studies suggest that BC has important climate effects causing short-term regional cooling but long-term global warming (Jacobson, 2002). Furthermore, due to the porosity of BC particles and their correspondingly large surface area, BC can adsorb a variety of chemicals that are present in combustion exhaust, including polycyclic aromatic hydrocarbons (PAHs), which are carcinogenic or mutagenic (Finlayson-Pitts and Pitts, 2000). Therefore, BC is a human health threat because of its ability to transport carcinogens to the lungs.

PAHs are a group of over 100 different compounds that are composed of two or more fused aromatic rings. They are a byproduct of incomplete combustion and are found in both gaseous and particulate phases. Some PAHs, such as benzo[a]pyrene, are carcinogenic (Denissenko et al., 1996), and a recent study has linked genetic abnormalities in newborns to their mothers’ exposure to PAHs during pregnancy (Boesky et al., 2005). Major urban sources of PAHs include residential wood combustion, tobacco smoking, cooking, and most of all, motor vehicle use. Contrary to the perception that diesel vehicles are the main vehicular source of PPAHs, light-duty gasoline vehicles have been found to be the most important source of PPAH emissions in some urban areas (Lobscheid and McKone, 2004). Concentrations of PPAHs along roadways in Mexico City are among the highest measured in the world (Marr et al., 2004), but quantitative estimates of their emissions are needed. Motor vehicles are expected to be the major source of PPAHs in Mexico City.

A variety of methods are used to measure motor vehicle emissions, including chassis dynamometer, tunnel, remote sensing, and mobile lab studies. A dynamometer is ideal for quantifying emissions from individual vehicles under a range of controlled driving conditions. However, dynamometer testing is practical for only a limited sample size, and thus it is difficult to represent the entire fleet and to estimate the emission inventory accurately. Tunnel and remote sensing studies can provide fleet-average emissions by sampling a larger number of vehicles, but because the measurements are conducted at a fixed site, they are usually restricted to certain driving conditions. Mobile laboratories, which are instrumented mobile platforms, including vans, trailers, boats, and airplanes (Kolb et al., 2004), can measure fresh tailpipe emissions while tailing vehicles. The advantage of the mobile lab is that it can measure emissions in real time over a wide range of real-world driving modes, e.g. idling, acceleration, cruising, and braking.

During the MCMA-2003 field campaign, a mobile lab was deployed in chasing mode, where individual vehicles were sampled in on-road chases, and in stationary mode, where the mobile lab remained fixed at a particular location for several days (Kolb et al., 2004). While analysis of individual chasing events provides emissions estimates for specific vehicles (Canagaratna et al., 2004), here we present a complementary approach which aims to estimate fleet-average emissions by employing all plumes measured by the laboratory as it is driven throughout the city. The assumption is that nearly all plumes measured on-road stem from vehicle exhaust, whether the mobile lab is chasing a specific vehicle or not. The objective of this research is to implement this approach for estimating fleet-average emissions of BC, PPAHs, and other species in Mexico City. Based on the measurement of emission ratios and by carbon balance, we characterize the distribution of emissions among the fleet and estimate the motor vehicle emission inventory.

2 Experimental

2.1 Mobile laboratory

A mobile lab designed and built by Aerodyne Research, Inc. (Billerica, MA) (Kolb et al., 2004) was deployed during the MCMA-2003 field campaign. It was equipped with a comprehensive suite of state-of-the-art fast response instruments, including an aethalometer (AE-16, Magee Scientific) for BC; a photoionization aerosol sensor (EcoChem PAS 2000CE) for particle-bound PAHs; a non-dispersive infrared (NDIR) unit (Li-Cor LI 6262) for carbon dioxide (CO$_2$); an Aerodyne tunable infrared laser differential absorption spectrometer (TILDAS) for nitrogen dioxide (NO$_2$) and formaldehyde (HCHO); an NDIR analyzer for carbon monoxide (CO); and a chemiluminescent analyzer (Thermo 42C) for nitrogen oxides (NO$_x$); a proton transfer reaction mass spectrometer (PTR-MS) for speciated volatile organic compounds (VOCs); and an aerosol photometer for PM$_{2.5}$ (TSI DustTrak 8520). Fast results for the aethalometer, which
Data collected by the mobile lab can be analyzed using two approaches: “microscopic,” which focuses on chases of individual vehicles and “macroscopic,” which considers all measurements as potential exhaust plumes. Building upon previous mobile lab studies in which fleet-average emission ratios were derived (Jimenez et al., 2000; Kolb et al., 2004), here we adopt the macroscopic approach, in which we identify many measurements representing a large number of vehicles in order to calculate fleet-average emissions. The laboratory continuously samples ambient air from an inlet at the front of the van, and it is always “seeing” exhaust plumes from the vehicles around it while driving through traffic, even when not actively chasing a particular vehicle.

In this analysis, we consider 75 h of data collected over 13 days of driving in all different directions from the field campaign’s base at the Universidad Autónoma de México (UAM) in Iztapalapa. As described in the following section, we analyze pollutant and meteorological measurements to develop criteria for the separation of fresh exhaust plumes from the on-road background.

2.3 Plume identification

The continuous measurements of pollutant concentrations represent three categories: 1) fresh tailpipe exhaust, 2) on-road background, and 3) exhaust contaminated by emissions from the mobile lab’s engine and generator, referred to as “self-sampling” in this paper. Our objective is to quantify emissions from on-road motor vehicles, so we wish to consider only those data points likely to reflect exhaust from other vehicles. Therefore, the identification and elimination of background and self-sampling measurements is an essential component of the calculation.

This approach follows the method suggested in previous analyses of fleet-average emissions using the mobile lab (Jimenez et al., 2000; Kolb et al., 2004). The identification of tailpipe exhaust and background sampling is based on concentrations of CO₂, which is a direct tracer of carbonaceous fuel combustion. As “baseline” is defined here as the minimum ambient pollutant concentration, the term “on-road background” refers to those data points that are not significantly above the “baseline.” “On-road background” points are therefore believed to represent measurements that are minimally influenced by nearby vehicles, as opposed to those heavily influenced by fresh vehicle exhaust. The baseline will change in space and time, as it is influenced by diurnal meteorological patterns and neighborhood-scale emissions. For example, the “baseline” in a neighborhood with numerous busy roads at a stagnant time of day will be higher than in a large park during a windy period.

The baseline is constructed by setting the fifth lowest value in a moving time window of 3 min, i.e., 180 1-s data points, as the baseline for all corresponding data points in the window. The fifth lowest value, rather than the lowest one, is used to account for outliers. We tested window widths of 1–15 min and found by visual inspection that a width of 3 min best balanced short- and long-term variability in the baseline for all instruments. The baseline is then smoothed by applying a binomial filter.

To determine a criterion for the separation of on-road background from fresh exhaust plume measurements, we analyze the CO₂ measurements during time periods when the van was parked away from traffic to find the natural variance of ambient CO₂ concentrations. We find that 95% of the data points lie within 42 ppm above the baseline. Therefore, we define 42 ppm (above baseline) as the threshold for potential exhaust plumes, as shown in Fig. 1. Points that exceed the threshold are considered exhaust plume measurements; those that fall below the threshold are considered on-road background measurements. Sensitivity tests reveal that a 10 ppm decrease in the threshold (to 32 ppm) results in a 4% decrease in calculated CO emission factors and that a 10 ppm increase (to 52 ppm) results in a 1% increase in CO emission factors. The relatively small change in emission factors, compared to the much larger change (24%) in the threshold, means that our results are not strongly dependent on the selection of a
particular threshold. The implications of not including some of the smaller plumes from vehicles are discussed later. We also considered other algorithms for plume identification, but their results did not match as well with the video record in identifying plumes. These alternate approaches included (1) requiring a certain threshold in the change in CO$_2$ concentration from one second to the next and (2) calculating ratios of the change in pollutant concentration to change in CO$_2$ from one second to the next.

“Self-sampling” is defined as sampling of air parcels that are contaminated by emissions from the mobile lab’s own tailpipe or its onboard generator. The generator was found to emit methanol (possibly from its coolant) which is not common in the exhaust of Mexico City’s vehicle fleet and which is therefore used as a tracer to identify self-sampling of the generator. When baseline-subtracted methanol concentrations exceed 30 ppb, a threshold determined by visual inspection of the data and video, the point is considered to be affected by self-sampling. Furthermore, the wind direction relative to the sampling inlet is also indicative of self-sampling. When the wind originates from behind the van or from the driver’s side where the van’s generator is located, the corresponding measurements are likely to be influenced by the van’s own emissions. In fact, when the wind direction is in the range of 180–300°, elevated methanol concentrations are often found, indicating self-sampling. We extend the self-sampling wind direction range to 90–300° to eliminate potential sampling of the mobile lab’s own tailpipe emissions from the rear. Overall, approximately 40% of all points are identified as self-sampling measurements.

After identifying potential exhaust plumes and self-sampling incidents, we construct a time series of markers that labels each 1-s data point as one of three types: valid exhaust measurements, on-road background measurements, and self-sampling. This time series, based on CO$_2$, wind direction, and methanol, is then applied universally to all other pollutant time series, e.g. CO, BC, PPAHs, to identify exhaust measurements. The assignment of sample types has been confirmed by examination of the video record at many brief intervals throughout the field campaign.

2.4 Emission factors

By performing a carbon balance on the fuel combustion process, one can relate the emissions of carbon-containing species in vehicle exhaust to fuel consumption. If $[C_f]$ refers to the carbon content originally in the fuel, $[CO_2]$, [CO], [VOC], [BC], and [PPAH] represent the carbonaceous species produced during combustion (mass of carbon per mass of fuel consumed), and $[C_0]$ accounts for all remaining carbon-containing species, such as fuel residues and other non-volatile organic compounds, the following mass balance should be observed:

$$[C_f] = [CO_2] + [CO] + [VOC] + [BC] + [PPAH] + [C_0]$$  \hspace{1cm} (1)

The sum of [BC], [PPAHs] and $[C_0]$ is very likely to be less than 0.1% of $C_f$ (Hansen and Rosen, 1990). We therefore simplify the equation to

$$[C_f] = [CO_2] + [CO] + [VOC]$$  \hspace{1cm} (2)

Then, the fuel-based emission factor can be calculated using the following equation:

$$E_p = \frac{\Delta[P]}{\Delta[CO_2] + \Delta[CO] + \Delta[VOC]} w_c$$  \hspace{1cm} (3)

where $E_p$ is the emission factor of pollutant P in grams of pollutant emitted per kilogram of fuel consumed; $\Delta[P]$ is the concentration of pollutant P above the baseline value, expressed in grams per cubic meter of air; $\Delta[CO_2]$, $\Delta[CO]$, and $\Delta[VOC]$ are the concentration increases of CO$_2$, CO, and VOC above their baseline values, expressed in grams of carbon per cubic meter of air; and $w_c$ is the mass fraction of carbon in the fuel. In this work, we compute the emission factors for the pollutants BC, PPAHs, CO, NO$_x$, benzene, total VOCs, and PM$_{2.5}$.

Due to the slower response times of some instruments (CO, NO$_x$, and PPAH), we integrate measurements over 10-s periods when at least 80% of the points represent exhaust plumes. The integration does not produce significantly different results for the species measured by instruments with true 1-s response times, such as benzene. For species measured by instruments with slower response times, the integration may result in a ~10% underestimate of emission factors due to the blending of “plume” and “background” values in the instruments’ response. This sensitivity test is not perfect because CO, with a slower response time, appears in the denominator of Eq. (3); however, its magnitude is very small compared to that of CO$_2$.

For estimation of fleet-average VOC emission factors, we use benzene as a surrogate for total VOCs because fast measurements of total VOCs are not available in the mobile
The VOC/benzene ratio, determined from 60 canister-based speciated VOC measurements collected during vehicle chases, is 65±28 by mass (Herndon et al., 2005\(^1\)). This result, derived from speciation values in ppbC, assumes that hydrocarbons dominate VOC mass and that a molecular mass of 14 for each carbon atom detected (each carbon carries two hydrogens) is representative of the mixture. We then apply this ratio to the results for benzene to estimate VOC emissions.

2.5 Emission inventory

By multiplying emission factors by the annual fuel sales in the MCMA, we can estimate the emission inventory, or the total emissions of each species. The area’s fuel sales were 6.82 and 1.50 billion liters per year of gasoline and diesel, respectively, in the year 2003 (Instituto Mexicano del Petróleo (IMP), 2001). We use an average fuel density of 760 g L\(^{-1}\) and carbon content of 0.85, based on a sales-weighted average of gasoline and diesel fuel properties (Gamas et al., 1999; Kirchstetter et al., 1999a; Schiffter et al., 2000).

The method is potentially biased by the fact that the number of data points representing different types of vehicles that were sampled may be disproportional to the actual distribution of vehicle types in the city. But it is reasonable to assume that the probability of measuring a certain vehicle type during an on-road experiment is consistent with the percentage of this specific type of vehicle in the city’s fleet. It is also reasonable to assume that large plumes, which are sampled in larger numbers due to their extended periods of high CO\(_2\) concentrations (and therefore greater fuel consumption), should in fact receive higher weight when considering their contribution to total emissions. In a fuel-based emission inventory for Los Angeles, the fleet-average emission factor was constructed by weighting emission factors specific to each model year by the model year’s fractional fuel consumption (Singer and Harley, 2000). To account for self-sampling by the mobile lab, we have excluded points containing methanol, but we may have inadvertently screened out other vehicles’ exhaust that contains methanol. At this point in time, such vehicles, mainly methanol-powered ones, are not expected to contribute significantly to Mexico City’s motor vehicle emission inventory. A final source of uncertainty stems from compositing fuel properties, i.e. carbon weight fraction and density, for use in Eq. (3) and emission inventory calculations. Carbon weight fractions differ by only 2% between gasoline and diesel fuel, but their densities differ by 13%. Our use of a sales-weighted density should minimize such error.

3 Results

3.1 Emission factors

Tables 1 and 2 summarize the emission factors of each species found in this study and in others that have taken place within Mexico and around the world. As shown in Table 2, CO and VOCs have the next highest emission factors among the carbonaceous species after CO\(_2\), representing 10% and 4% respectively of carbon in the exhaust on average, while BC and PPAHs (Table 1) rank among the lowest emission factors and account for 0.03% and <0.001% of carbon in the exhaust, which is consistent with our earlier assumption for the simplified mass balance. Also note that the standard deviations (\(s_x\)) are generally large because of the skewed

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A remote sensing study that reports the mean and standard deviation across nine days of sampling at five locations (Beaton et al., 1992).

A remote sensing study that reports the mean and standard deviation across four site means (Bishop et al., 1997).

Hydrocarbons only. These values do not include the factor-of-two correction to account for varying absorption by different species (Singer and Harley, 2000).

Not available.

A remote sensing study that reports the mean and 95% confidence intervals across 12 site means (Schifter et al., 2005).

Table 2. Fuel-Based Emission Factors (g kg\(^{-1}\)) from on-road motor vehicles measured in this study and in remote sensing studies in Mexico.

<table>
<thead>
<tr>
<th>(g kg(^{-1}))</th>
<th>Vehicle types</th>
<th>CO</th>
<th>Benzene</th>
<th>VOC</th>
<th>NO(_x)</th>
<th>PM(_{2.5})</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mexico City 2003(^b)</td>
<td>all</td>
<td>190±160(^a)</td>
<td>0.60±0.32</td>
<td>38±27</td>
<td>19±12(^c)</td>
<td>0.7±1.4(^d)</td>
</tr>
<tr>
<td>Mexico City 2000(^e)</td>
<td>light-duty</td>
<td>155±18</td>
<td>N/A(^f)</td>
<td>18±3(^g)</td>
<td>21±5</td>
<td>N/A</td>
</tr>
<tr>
<td>Monterrey 1995(^b)</td>
<td>light-duty</td>
<td>205±23</td>
<td>N/A</td>
<td>12±3(^g)</td>
<td>30±8</td>
<td>N/A</td>
</tr>
<tr>
<td>Mexico City 1991(^i)</td>
<td>light-duty</td>
<td>475±200</td>
<td>N/A</td>
<td>96±29(^g)</td>
<td>N/A</td>
<td>N/A</td>
</tr>
</tbody>
</table>

\(^a\) NO\(_x\) is reported as NO\(_2\).

\(^b\) Mean and standard deviation across all individual measurements.

\(^c\) Values reported for species measured by analyzers with slower response times may be underestimated by up to 10%, as discussed in the text.

\(^d\) This estimate does not include the additional uncertainty imposed by the limitations of the PM\(_{2.5}\) measurement method, as discussed in the text.

\(^e\) A remote sensing study that reports the mean and 95% confidence intervals across 12 site means (Schifter et al., 2005).

\(^f\) Not available.

\(^g\) Hydrocarbons only. These values do not include the factor-of-two correction to account for varying absorption by different species (Singer and Harley, 2000).

\(^h\) A remote sensing study that reports the mean and standard deviation across four site means (Bishop et al., 1997).

\(^i\) A remote sensing study that reports the mean and standard deviation across nine days of sampling at five locations (Beaton et al., 1992).

distributions, but the large number of measurements (~2300 valid 10-s periods) results in a much smaller standard error and narrower confidence interval.

Table 1 lists the values of emission factors of BC and/or PPAHs determined in previous studies in the US and Switzerland. While the emission factors measured in Mexico City represent a fleet-average that incorporates both gasoline- and diesel-powered vehicles, many of the other studies provide separate emission factors for light-duty (mainly gasoline-powered) and heavy-duty (mainly diesel-powered) vehicles. The emission factors measured in Mexico City, shown at the bottom of the table, fall within the range of measurements found in the other studies.

As shown in Fig. 2, the distributions of emission factors of different species are skewed to varying degrees. The distributions of those species with modes near zero, such as BC, PPAHs, and PM\(_{2.5}\), are dramatically skewed by a small number of high-value points, while species like NO\(_x\) and benzene are more normally distributed. The skewness coefficient, a measure of the symmetry of a distribution where 0 is perfectly symmetric, of BC and PM\(_{2.5}\) emission factors is 5–6; and that of the other species is <2 in all cases. The few data points with very high values considerably affect the means of BC, PPAHs, CO, and PM\(_{2.5}\). The implications of the different distribution patterns are discussed in the following section.

3.2 Emission inventory

Table 3 compares our results to the Mexican government’s official motor vehicle emission inventory for the year 2002, the most recent year for which such estimates are available. For NO\(_x\) and VOCs, our estimates represent lower limits due to the instruments’ slower response times, as discussed previously. The Secretaría del Medio Ambiente y Recursos Naturales, the Mexican equivalent of the US Environmental Protection Agency (EPA), has been producing inventories since 1996 using a customized version of EPA’s MOBILE model and light-duty gasoline emission factors measured by the Instituto Mexicano del Petróleo. Comparison of the official inventory against independent estimates can provide a measure of uncertainty in the inventory.

The reasonably good agreement between the two sets of estimates adds confidence to them. As shown in Fig. 3, our estimates of Mexico City’s motor vehicle emissions in 2003 for CO, NO\(_x\), VOCs, and PM\(_{2.5}\) are 38% lower, 23% lower, 27% higher, and 25% higher, respectively, than found in the government’s official motor vehicle emission inventory for the year 2002. For reference, it is not uncommon for such estimates to disagree by factors of two (200%) or more (Fine et al., 2003; Russell and Dennis, 2000), or even by factors of up to ten (Ryerson et al., 2003), so the agreement found here is relatively good. Emission inventory models such as MOBILE have underestimated VOC and CO emissions in the past (Arriaga-Colina et al., 2004; Sawyer et al., 2000), yet our results and those of another study using fuel-based methods (Schifter et al., 2005) suggest the opposite.

4 Discussion

4.1 Emission factors

As shown in Table 2, the emission factors of CO, NO\(_x\), and VOCs measured in this study are similar to those found in a remote sensing study in Mexico City in 2000 (Schifter et
Fig. 2. Normalized distributions and averages of emission factors of BC, PPAH, CO, NOx, benzene, and PM2.5. Note that the range of the y-axis varies.

Our results are substantially lower than found in a remote sensing study in Mexico City in 1991 (Beaton et al., 1992). Our mean CO emission factor is 23% higher than the year 2000 measurement. The difference may stem from the mobile lab’s ability to capture a wider range of engine loads compared to remote sensing, which is typically limited to sites with a single lane of traffic and requires uphill grades or locations with moderate accelerations to capture a signal. The NOx emission factors agree well. If the remote sensing VOC measurement from 2000 is scaled upward by a factor of two, as suggested to account for varying absorptions by different VOCs (Singer et al., 1998), then the corrected value of 36 g kg\(^{-1}\) is essentially equivalent to our estimate.

Table 2 also shows that over a 12-year period between 1991 and 2003, CO and VOC emission factors fell by 60% and 80% (again applying the factor-of-two correction for the 1991 VOC measurement), respectively, if we assume that remote sensing and the mobile lab provide equivalent results. At the very least, we can assert that emission factors of these two pollutants have fallen significantly. The introduction of catalytic converters in 1991 likely played a large role in these reductions, with the effect slowly propagating through the fleet as older cars were gradually replaced. With respect to total emissions, reductions in emission factors are partially offset by growth in fuel consumption over this period, such that the corresponding change in the emission inventory is not expected to be as large.

Mexico City’s emission factors are considerably higher than those measured in the US. Because gasoline-powered vehicles are the predominant source of vehicular CO and VOC emissions (Sawyer et al., 2000), we focus on comparing Mexico City’s emission factors of these two species to measurements of light-duty gasoline-powered vehicles in the US. Remote sensing measurements in Denver found a light-duty fleet-average CO emission factor of 65 g kg\(^{-1}\) in 1999–2000 (Pokharel et al., 2002). In the Caldecott Tunnel in the San Francisco Bay Area in 2001, light-duty CO emission factors were 34–81 g kg\(^{-1}\) over a range of engine loads with vehicle specific powers of −5 to 18 W kg\(^{-1}\) (Kean et al., 2003).
Table 3. Comparison of the motor vehicle emission inventory in Mexico City.

<table>
<thead>
<tr>
<th>Pollutant (metric tons yr⁻¹)</th>
<th>This study⁹</th>
<th>Official inventory or alternative approach</th>
</tr>
</thead>
<tbody>
<tr>
<td>BC</td>
<td>1700±200</td>
<td>2300±100⁹</td>
</tr>
<tr>
<td>PPAHs</td>
<td>57±6b</td>
<td>27±1⁹</td>
</tr>
<tr>
<td>CO</td>
<td>1190,000±40,000⁹</td>
<td>1927±101</td>
</tr>
<tr>
<td>NOₓd</td>
<td>120,000±30,000⁹</td>
<td>156,311</td>
</tr>
<tr>
<td>VOC</td>
<td>240,000±50,000</td>
<td>188,530</td>
</tr>
<tr>
<td>PM₂,₅</td>
<td>4400±400⁹</td>
<td>3518</td>
</tr>
</tbody>
</table>

⁹ Range shows 95% confidence interval, and results from this study do not include cold starts.

Values reported for species measured by analyzers with slower response times may be underestimated by up to 10%, as discussed in the text.

The government produces estimates of CO, VOC, NOₓ, and PM₂,₅ emissions (Comisión Ambiental Metropolitana, 2004) but does not estimate emissions of BC or PPAH. To develop alternative estimates for these species, we multiplied the ambient ratios \( \Delta[\text{BC}]/\Delta[\text{CO}] \) and \( \Delta[\text{PPAH}]/\Delta[\text{CO}] \) measured during the field campaign, by our estimate of total CO emissions.

d ⁹ Reported as NO₂.

e ⁹ This estimate does not include the additional uncertainty imposed by the limitations of the PM₂,₅ measurement method, as discussed in the text.

The emission factor of 190 g kg⁻¹ measured in Mexico City is 2.4–5.6 times higher than found in the U.S. The difference for VOC emission factors is even larger. Light-duty non-methane hydrocarbon (NMHC) emission factors in the Caldecott Tunnel were 1.3 and 4.3 g kg⁻¹ for steady uphill versus downhill driving, respectively. Oxygenated compounds comprise ~12% of total VOCs in this exhaust (Kirchstetter et al., 1999b), so the VOC emission factors can be estimated to range between 1.5 and 4.9 g kg⁻¹. Mexico City’s emission factor of 38 g kg⁻¹ measured in 2003 is at least seven times higher. A previous comparison using the mobile lab found that the ratio of formaldehyde to CO₂ was seven times higher in Mexico City than in Boston (Kolb et al., 2004).

Because our NOₓ emission factors represent a fleet-average that includes both light- and heavy-duty vehicles, a direct comparison to measurements in the US is not possible. Remote sensing and tunnel studies have reported light-duty NOₓ emission factors of 7 g kg⁻¹ in the U.S. in recent years (Harley et al., 2005; Pokharel et al., 2002). The average heavy-duty NOₓ emission factor in the US is 39 g kg⁻¹ (Marr et al., 2002) and has not changed significantly over the past 20 years (Yanowitz et al., 2000). Mexico City’s combined NOₓ emission factor of 19 g kg⁻¹ falls in between light- and heavy-duty values found in the U.S.

The skewness of the emission factor distributions shown in Fig. 2 may indicate the degree to which a pollutant is mainly emitted by certain types of vehicles that make up a small fraction of the fleet. BC is expected to be emitted mainly by diesel vehicles, while the other pollutants are also emitted by gasoline-powered vehicles, which make up the majority of the fleet. The distributions and their skewness may also be indicative of different driving modes.

The discrepancies between the mean and the mode of the distributions shown in Fig. 2 illustrate the large impact that a very small fraction of the vehicle fleet may have on the average emission factor. In fact, if the highest 20% of the points are eliminated, the average emission factors of BC, PPAHs, and PM₂,₅ can be reduced by ~50%, which corresponds to comparable reductions in the emission inventory. On the other hand, species like NOₓ and benzene are not skewed as much by “super polluters.” Eliminating the highest 20% of values will only result in a 24% and 23% reduction in their emissions, respectively. For CO, whose emission factors are moderately skewed, eliminating the highest 20% of values reduces the overall average by 36%. These distribution...
patterns suggest that policies focusing on a small fraction of the vehicle fleet, i.e. targeting “super polluters,” will have the greatest effect on BC and PPAH emissions and less of an effect on NO\textsubscript{x} and benzene (or VOCs).

4.2 Emission inventory

Although we can conclude that the mobile lab and government model give approximately similar results, at least in comparison to previous studies, two sources of uncertainty prevent us from making stronger quantitative conclusions about the accuracy of the inventory. First, cold-start emissions are not captured in our on-road experiments and are therefore not included in our estimates. Second, using the VOC/benzene ratio to estimate VOC emissions adds uncertainty, as we are assuming that the VOC canisters contain exhaust from different sources in proportions that are representative of their contribution to the total emission inventory. The overall uncertainty in VOC emissions can be determined quantitatively in the future when a method for fast measurement of total VOCs becomes available in the mobile lab.

We also compare our estimates of the total emissions of BC and PPAHs to results determined using an alternative approach, in which ambient BC/CO and PPAH/CO ratios are multiplied by the total CO emissions in the city (Table 3). The ratio of BC/CO measured at the supersite in Iztapalapa during the MCMA-2003 field campaign was 1.63±0.02 \mu g m\textsuperscript{-3} ppm\textsuperscript{-1} (r\textsuperscript{2}=0.64), or 1.89±0.03×10\textsuperscript{-3} g BC g\textsuperscript{-1} CO. The resulting estimate of BC emissions, shown in Table 3, is 35% higher than our fuel-based result, which is consistent with the common understanding that ambient measurements will detect BC emissions from sources besides on-road motor vehicles. Likewise, the ambient PPAH/CO ratio was 19.5±0.2 ng m\textsuperscript{-3} ppm\textsuperscript{-1} (r\textsuperscript{2}=0.71), or 2.27±0.03×10\textsuperscript{-5} g PPAH g\textsuperscript{-1} CO, but the alternative estimate is 53% lower than the result from the mobile lab. We believe the difference is largely due to particle aging or coating that occurs during the transport from vehicle tailpipes to the ambient monitoring site. Particle aging, through coagulation and coating by secondary aerosol, causes a large portion of the surface-adsorbed PAHs to become undetectable to the photoionization aerosol sensor, which is sensitive only to PAHs on particle surfaces.

The advantage of the mobile lab is its ability to provide a comprehensive estimate of the vehicle fleet’s emissions, taking into consideration a variety of vehicle types under the full range of real-world driving conditions. In future work, we suggest using additional statistical analyses to identify the different types of engines associated with each exhaust plume point. Principal component analysis could be used to study the expected covariance between, for example, high NO\textsubscript{x} and high BC emissions that might be associated with diesel vehicles.

5 Conclusions

We have used a mobile lab to measure fleet-average emission factors of BC, PPAHs, and other species in Mexico City. Out of 75 h of sampling, the method has identified ~30 000 data points corresponding to exhaust plumes. Motor vehicles are estimated to emit 1700±200 metric tons BC, 57±6 tons PPAHs, 190 000±40 000 tons CO, 120 000±300 tons NO\textsubscript{x}, 240 000±50 000 tons VOCs, and 4400±400 tons PM\textsubscript{2.5} per year. The estimates for CO, NO\textsubscript{x}, and PPAHs may be low by ~10% due to the slower response time of analyzers used to measure these species. The distributions of the emission factors are skewed to varying degrees. As a result, the total emissions of BC and PPAHs can be reduced by approximately 50% if the highest 20% of data points are removed, but “super polluters” are less influential on NO\textsubscript{x} and VOC emissions. A fast measurement technique for total VOCs is desired in order to more directly estimate VOC emissions using this method. This method can be combined with manual analysis of chase events, so that the emissions of individual vehicles or vehicle classes can be studied while self-sampling and on-road background measurements are conveniently detected and separated from those of real exhaust plumes.

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