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## 8 **Abstract**

9 East African countries face an increasing threat from poor air quality, stemming from rapid  
10 urbanisation, population growth and a steep rise in fuel use and motorization rates. With few air  
11 quality monitoring systems available, **this study provides much needed** high temporal resolution data  
12 to investigate the concentrations of particulate matter (PM) air pollution in Kenya. Calibrated low cost  
13 optical particle counters (OPCs) were deployed in Kenya in three locations: two in the capital of Nairobi  
14 and one in a rural location in the outskirts of Nanyuki, which is upwind of Nairobi. The two Nairobi  
15 sites consist of an urban background site and a roadside site. The instruments were composed of an  
16 AlphaSense OPC-N2 optical particle counter (OPC) ran with a raspberry pi low cost microcomputer,  
17 packaged in a weather proof box. Measurements were conducted over a two-month period (February  
18 – March 2017) with an intensive study period when all measurements were active at all sites lasting  
19 two weeks. When collocated, the three OPC-N2 instruments demonstrated good inter-instrument  
20 precision with a coefficient of variance of  $8.8 \pm 2.0\%$  **in the fine particle fraction (PM<sub>2.5</sub>)**. The low cost  
21 sensors had an absolute PM mass concentration calibration using a collocated gravimetric  
22 measurement at the urban background site in Nairobi.

23 The mean daily PM<sub>1</sub> mass concentration measured at the urban roadside, urban background and rural  
24 background sites were 23.9, 16.1, 8.8  $\mu\text{g m}^{-3}$ . The mean daily PM<sub>2.5</sub> mass concentration measured at  
25 the urban roadside, urban background and rural background sites were 36.6, 24.8, 13.0  $\mu\text{g m}^{-3}$ . The  
26 mean daily PM<sub>10</sub> mass concentration measured at the urban roadside, urban background and rural  
27 background sites were 93.7, 53.0, 19.5  $\mu\text{g m}^{-3}$ . The urban measurements in Nairobi showed that  
28 particulate matter concentrations regularly exceed WHO guidelines in both the PM<sub>10</sub> and PM<sub>2.5</sub> size  
29 ranges. Following a 'Lenschow' type approach we can estimate the urban and roadside increments  
30 that are applicable to Nairobi **(Lenschow et al., 2001)**. **The median urban increment is 33.1  $\mu\text{g m}^{-3}$  and**  
31 **the median roadside increment is 43.3  $\mu\text{g m}^{-3}$  for PM<sub>2.5</sub>. For PM<sub>1</sub>, the median urban increment is 4.7**

1  $\mu\text{g m}^{-3}$  and the median roadside increment is  $12.6 \mu\text{g m}^{-3}$ . These increments highlight the importance  
2 of both the urban and roadside increments to urban air pollution in Nairobi.  
3 A clear diurnal behaviour in PM mass concentration was observed at both urban sites, which peaks  
4 during the morning and evening Nairobi rush hours; this was consistent with the high measured  
5 roadside increment indicating that vehicular traffic is a dominant source of particulate matter in the  
6 city, accounting for approximately 48.1, 47.5, and 57.2% of the total particulate matter loading in the  
7  $\text{PM}_{10}$ ,  $\text{PM}_{2.5}$  and  $\text{PM}_1$  size ranges, respectively. Collocated meteorological measurements at the urban  
8 sites were collected, allowing for an understanding of the location of major sources of particulate  
9 matter at the two sites. The potential problems of using low cost sensors for PM measurement without  
10 gravimetric calibration available at all sites are discussed.  
11 This study shows that calibrated low cost sensors can be used successfully to measure air pollution in  
12 cities like Nairobi. It demonstrates that low cost sensors could be used to create an affordable and  
13 reliable network to monitor air quality in cities.

## 14 **1. Introduction**

15 Recently, the Lancet Commission on pollution and health estimated that in 2015, air pollution led to  
16 the premature deaths of over nine million people globally, and contributed to over one in four deaths  
17 in severely affected countries (Landrigan et al., 2017). Typically, urban air pollution is higher in low  
18 and middle-income countries (LMICs) compared to further developed countries. Hence, the associated  
19 risk of air pollution to health is typically higher in LMICs, with over 92% of global pollution related  
20 deaths occurring in these countries. Within LMICs, health inequalities in urban areas contribute to an  
21 increased exposure to air pollution that faces those that live, work, socialise and commute to highly  
22 urbanised areas which typically have a substantially higher concentration of air pollutants. Despite the  
23 extensive links between air pollutants and human health, environmental degradation and the  
24 economy, air pollution is as of yet still under-researched in many LMICs. Due to a lack of long term air  
25 quality monitoring in many LMICs, the concentrations and sources of air pollution are poorly  
26 understood.

27 Airborne particulate matter (PM) is a major environmental risk factor with well-documented short and  
28 long-term effects on human mortality and morbidity (Thurston et al., 2016). Long term side effects to  
29 air pollution exposure include asthma, chronic pulmonary disease (COPD), pulmonary fibrosis, cancer,  
30 type-2 diabetes, neurodegenerative diseases, obesity and other conditions (Ferranti et al., 2017;  
31 Landrigan et al., 2017). People who are already susceptible to underlying respiratory disease,  
32 pneumonia, influenza or asthma can see a worsening of their symptoms and illnesses through short  
33 term exposure to air pollution (Wan Mahiyuddin et al., 2013). The size of PM is correlated with their

1 health impacts, with smaller particles typically having more significant health implications (Meng et  
2 al., 2013).  $PM_{1}$ ,  $PM_{2.5}$  and  $PM_{10}$  are particulate matter with aerodynamic diameters less than 1, 2.5  
3 and 10  $\mu m$ , respectively. The World Health Organization (WHO) recommends that  $PM_{2.5}$  and  $PM_{10}$  daily  
4 mass concentrations should not exceed 25 and 50  $\mu g/m^3$ , respectively; and that annual mass  
5 concentrations do not exceed 10 and 20  $\mu g/m^3$ , respectively (WHO, 2006). At present, the WHO or  
6 other regulatory bodies do not provide recommendations of the mass concentrations of  $PM_{1}$ .  $PM_{1}$  can  
7 remain suspended in air for much longer than coarser particulate matter, as well as penetrating  
8 deeper into the lungs leading to local pulmonary, systematic inflammation (Pateraki et al., 2014). Due  
9 to the smaller size,  $PM_{1}$  has a higher surface to mass ratio, containing a harmful amount of potentially  
10 toxic anthropogenic constituents which could lead to health impacts such as respiratory disease, heart  
11 disease and lung cancer (Trippetta et al., 2016). Many studies still focus on  $PM_{10}$  and  $PM_{2.5}$  even though  
12 smaller particulates pose greater health impacts (Tsiouri et al., 2015). Beyond  $PM_{1}$ , ultra-fine particles  
13 (<100 nm) are of such a small size they can be translocated to the central nervous system via the blood  
14 to brain barrier or the olfactory bulb (Chen et al., 2016). There are no air quality regulations of  $PM_{1}$  or  
15 ultra-fine particles due to the paucity of data either within environmental science or public health.

16 Worldwide, road traffic is a dominant source of urban PM accounting for 5-80% of PM mass, with the  
17 precise amount being dependent upon several factors including time, location, and vehicle fleet, as  
18 reviewed by Pant and Harrison (2013). Vehicle derived PM is directly associated with negative health  
19 outcomes (Fan et al., 2006;HEI, 2010). Emissions are due both to exhaust pipe emissions and non-  
20 exhaust pipe emissions. Exhaust emissions result from the combustion of fuel, predominantly petrol  
21 and diesel, and oil and other lubricants. Non-exhaust emissions come either from the resuspension  
22 of road dust through wind or vehicle induced wind shear, or from the wear and tear of vehicle parts  
23 including the brakes, tyres and clutch. Resuspension of dust is particularly important on non-paved  
24 roads of which there are an abundance in Nairobi. Typically, non-exhaust emissions are in the coarse  
25 PM size fraction (PM in the size range 2.5-10  $\mu m$  aerodynamic diameter), whereas exhaust emissions  
26 are in the fine PM size fraction ( $PM_{2.5}$ ) (Thorpe et al., 2007;Kam et al., 2012). However, it is noted that  
27 the papers which reference vehicle PM size distributions according to the emission of non-exhaust  
28 sources have typically been conducted in either the US or European studies and not in Africa, where  
29 non-paved road sources represent a much higher fraction of road surface type. The precise size of  
30 vehicular derived PM is dependent on several factors: vehicle fleet characteristics (e.g. weight and  
31 size), road type and level of maintenance and meteorological conditions (Beddows et al., 2009;Hays  
32 et al., 2011).

1 In many LMIC cities, urbanization, population, fuel use and motorization rates are all increasing rapidly  
2 and increases in air pollution are associated with these trends (Mitlin and Satterhwaite, 2013;Ochieng  
3 et al., 2017). In particular, vehicular traffic is fast on the rise, with associated congestion on the road  
4 networks, which can contribute as much as 90% of air pollution in urban environments (UNEP, 2005).  
5 Nairobi is the capital city of Kenya and is showing these trends. In particular, the city population has  
6 increased dramatically, since 1999 to 2015 it has risen by 83%, and is projected to increase to 7.14  
7 million by 2030 (Rajé et al., 2017). Similarly, motorization rates are increasing, between 2008 and  
8 2012, the number of motor- and auto-cycles in Kenya grew by 368% with the number of overall  
9 registered vehicles increasing by 77% (Rajé et al., 2017). Considering this extensive increase in the  
10 vehicle fleet, limited roadway infrastructure and high congestion within the city, pollution hotspots  
11 are created leading to personal exposure levels much higher than that encountered throughout the  
12 rest of the city (van Vliet and Kinney, 2007).

13 To be able to reduce air pollution, it is helpful to be able to measure it, so reduction efforts can be  
14 assessed. Many LMIC countries have insufficient monitoring networks through which to measure air  
15 quality. In particular, long term high resolution data is required for such cities which are vulnerable to  
16 air pollution. Nairobi is in the vanguard of air pollution measurements for Sub-Saharan Africa but lacks  
17 continuous long term calibrated measurements of PM and other air pollutants (Petkova et al., 2013).  
18 A discussion of the relevant measurements in Nairobi is given in the next section. One of the  
19 constraints to making measurements is the high cost of research grade air quality monitoring  
20 equipment with appropriate calibration and certification. Low cost sensors offer the potential for  
21 dramatically reducing equipment costs by orders of magnitude, making the monitoring of air quality  
22 more accessible and attainable in LMIC countries (Lewis et al., 2016;Rai et al., 2017).

23 In this paper, the use of low cost sensors for measurement of PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> in Nairobi is  
24 detailed. We have previously assessed the same low cost sensors in the UK (Crilley et al., 2018). The  
25 sensors are calibrated using a standardised gravimetric approach. PM is measured in three locations:  
26 an urban roadside site, an urban background site and a rural background site. Comparison of  
27 simultaneous measurements at the three sites allows for the estimation of an urban increment and  
28 roadside increment in PM following a 'Lenschow' type approach which allows for simple modelling of  
29 city air pollution based on the urban and roadside increments in air pollution (Lenschow et al., 2001).  
30 The variation of measured PM with measured meteorological data is also discussed. Finally, we discuss  
31 the implications of using low cost sensors in Nairobi and LMIC countries in general.

## 1 **2. Previous PM measurements in Nairobi**

2 In general, long term air quality monitoring in Sub-Saharan Africa (SSA) is rare. Correspondingly, there  
3 are only limited PM data sets for East African urban areas; where data does exist estimated  
4 concentrations for PM<sub>2.5</sub> concentrations are approximately 100 µg/m<sup>3</sup> compared to <20 µg/m<sup>3</sup> in most  
5 European and North American cities (Brauer et al., 2012). This indicates that urban PM air pollution  
6 in East Africa could be a significant health concern.

7  
8 In Nairobi, there have been numerous short term measurements of PM over the last decade (Brauer  
9 et al., 2012; Kinney et al., 2011; Ngo et al., 2015; Egondi et al., 2016; Gaita et al., 2016) with only one  
10 long term continuous measurement (Gaita et al., 2014). To date, most measurements have used  
11 gravimetric measurement methodologies to record PM mass concentration in the PM<sub>2.5</sub> and PM<sub>10</sub> size  
12 fractions. Most measurements indicate PM concentrations in Nairobi regularly exceed the WHO  
13 guidelines. At present, there is only one publication in the scientific literature describing the use of  
14 low cost sensors in the measurement of PM (de Souza et al., 2017) which monitored air quality in  
15 Nairobi at six sites from May 2016 to January 2017. Using AlphaSense OPC-N2's, the authors measured  
16 PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> as well as NO<sub>2</sub>, NO and SO<sub>2</sub>.

17 The study collected PM concentrations at six schools within Nairobi. It reported a PM<sub>2.5</sub> concentration  
18 range between 11 and 21 µg/m<sup>3</sup>, and a range of 26 to 59 µg/m<sup>3</sup> for PM<sub>10</sub>. The PM concentrations  
19 measured during the de Souza study are noticeably lower than of this study for both size fractions. It  
20 is worthy of note that the de Souza study collected measurements from May 2016 to January 2017,  
21 whereas this campaign took place from February to April 2017; the local meteorology may have  
22 influenced the discrepancies seen in both recorded PM concentrations. Additionally, the study did not  
23 calibrate the monitors, which leads to questions about absolute concentrations and interference from  
24 other environmental dependencies (Lewis and Edwards, 2016). The collected data from the study  
25 appeared noisy, with the authors stating they could not separate the signal from the noise without  
26 having access to an air quality measuring reference instrument (they recorded peaks at over 1000  
27 µg/m<sup>3</sup>). Despite the limitations, it provides a useful comparison to this calibrated study.

28  
29 The paucity of long term calibrated measurements has hindered the understanding of long term  
30 trends and the influence of seasonal variations in meteorology and other factors. Most published data  
31 provide daily averages of PM mass; the lack of higher temporal resolution data precludes the  
32 generation of diurnal data which can be useful for identifying individual sources of PM, in particular,  
33 vehicular PM which typically tracks traffic and hence peaks during rush hours.

34

1 The longest record of PM concentration in Nairobi is detailed in Gaita et al. (2014). In this work, the  
2 authors performed daily measurements of PM<sub>2.5</sub> at an urban background and suburban site over a  
3 two-year period from May 2008 to April 2010 using polycarbonate filters in cyclone sampler (Casella  
4 from Bedford, UK). They reported a concentration range of 3 µg/m<sup>3</sup> to 53 µg/m<sup>3</sup> at the urban  
5 background site, with an overall mean of 21 µg/m<sup>3</sup> which exceeds the annual WHO limit of 10 µg/m<sup>3</sup>  
6 by a factor of two. The average concentrations of PM<sub>2.5</sub> at both sites were found to be 21 ± 9.5 and 13  
7 ± 7.3 µg/m<sup>3</sup>, respectively. Chemical composition measurements of the filter samples allowed source  
8 apportionment, via positive matrix factorization, to be carried out. The analysis suggested that five  
9 major source factors contribute to Nairobi PM<sub>2.5</sub>: traffic, mineral dust, industry, combustion and a  
10 mixed factor. The dominant source factors were mineral dust and traffic which accounted for 74% of  
11 the particle mass.

12

13 As an update to this study, Gaita et al. (2016) conducted a study on the characterization and size-  
14 fractionation of particulate matter and deposition fraction in the human respiratory system in Nairobi  
15 using measurements taken in August and September 2007, obtained at the University of Nairobi site.  
16 Based on the findings, the concentration levels of airborne particulate matter sampled at the urban  
17 background site during the period was found to range between 1 µg/m<sup>3</sup> and 78 µg/m<sup>3</sup>. The average  
18 PM<sub>2.5</sub> concentration at the site over the entire sampling period was 9.8±8.5 µg/m<sup>3</sup>.

19 A densely populated urban area with associated heavy local traffic within Nairobi largely contributes  
20 to the city's air pollution build up. Kinney et al., (2011) investigated the impact of vehicular emissions  
21 in Nairobi on the concentration of PM<sub>2.5</sub>, observing a substantial range between 58 µg/m<sup>3</sup> and 98  
22 µg/m<sup>3</sup> across an 11-hour personal exposure along busy roadways and roundabouts. The range could  
23 be estimated to be between 45 and 85 µg/m<sup>3</sup> for a 24 h sampling due to pollutant dispersion at night.  
24 In addition, the study reported a decrease in horizontal dispersion measurements of PM<sub>2.5</sub> from 128.7  
25 µg/m<sup>3</sup> to 18.7 µg/m<sup>3</sup> over 100 m downwind of a major intersection in Nairobi. A vertical dispersion  
26 from a street level to a third-floor rooftop in the Central Business District (CBD) showed a decrease in  
27 PM<sub>2.5</sub> concentration from 119.5 µg/m<sup>3</sup> to 42.8 µg/m<sup>3</sup>. This study clearly highlights that the PM  
28 concentration in Nairobi varies considerably over both time and space, which has significant  
29 implications for human exposure, see discussion.

30

31 Another study by Ngo et al., (2015) affirmed the contribution of anthropogenic activities on the quality  
32 of air in Nairobi. In their study, Teflon filters in PM<sub>2.5</sub> samplers (BGI model 400) were used between 2<sup>nd</sup>  
33 August and 18<sup>th</sup> August 2011 and high concentrations of PM<sub>2.5</sub> exposure levels among different groups  
34 in Nairobi were reported. According to the study, bus drivers in Nairobi city were exposed to about

1 103  $\mu\text{g}/\text{m}^3$  while those in informal settlements, such as Mathare, reporting exposure levels of about  
2 62.7  $\mu\text{g}/\text{m}^3$ , an indication that urgent measures needed to be taken to mitigate the impact of air  
3 pollution in the city.

4  
5 The severity of air pollution in urban centres in SSA is typically even higher in the informal settlements  
6 (slums), where acute respiratory tract infections and bronchitis are among the most frequent medical  
7 diagnoses (Gulis et al., 2004). Egondi et al., (2016), in their study on air pollution in two informal  
8 settlements in Nairobi: Korogocho and Viwandani, reported higher concentration levels of  $\text{PM}_{2.5}$  in the  
9 two slums. Optical counters (TSI DustTrak II model 3530) were used in the study and observed average  
10 concentration levels of  $\text{PM}_{2.5}$  in Korogocho slum, lying west of Dandora, Nairobi's biggest dumping  
11 ground, were the highest at 166  $\mu\text{g}/\text{m}^3$  and Viwandani, situated North of Nairobi recorded 67  $\mu\text{g}/\text{m}^3$ .

### 12 **3. Methodology**

#### 13 **3.1. Site locations**

14 This study utilised three field sites in Kenya, see Figure 1. Two sites were in Nairobi which is the capital  
15 of Kenya, covering an area of ca. 696 sq. kilometres and home to approximately 3.5 million residents  
16 according to a World Population review conducted in 2016 (Kenya Population, 2016), making it the  
17 second most populated city in East Africa after Dar es Salaam, Tanzania. In addition to Nairobi's  
18 longstanding popularity as a travel destination, due to its safari and other holiday resorts, the city also  
19 acts as East Africa's diplomatic, financial and communication capital (Rajé et al., 2017).

20 Its geographical location is at approximately 1.29° S and 36.82° E. The highest elevation point in the  
21 city is at an altitude of 1663m above the ground. As discussed in the introduction, Nairobi is  
22 undergoing rapid increases in population and motorization both of which will likely lead to greater PM  
23 pollution in the absence of any efforts of mitigation against the pollution. Other significant  
24 infrastructure projects such as major road building are currently being undertaken, which will also  
25 likely lead to increased PM loadings. Within Nairobi, the two field sites represent an urban  
26 background location and an urban roadside location. The other site, a rural background site is located  
27 on the outskirts of Nanyuki, a town that is located at an approximate aerial distance of 147 km to the  
28 north (NNE) of Nairobi and 240 km by road. The sensor boxes were placed in locations free from  
29 obstacles, at the three measurement sites, allowing for 360 degrees of air flow. A description of the  
30 meteorological conditions is provided in the supplementary material.

1 **Site 1: American Wing, University of Nairobi, Nairobi (urban background site)**

2 The first site for data collection in Nairobi was at the American Wing building located in the University  
3 of Nairobi, standing at an elevation of 17 m above ground level. Air flow at the site was free from any  
4 obstruction as the OPC's were located at an elevated point above the ground. The nearest road is  
5 Harry Thuku road which has very few on-road vehicles (no heavy trucks) and it leads to Fairmont  
6 Norfolk and Boulevard hotels, and Kijabe Street. Its level is far below the site and the only influence  
7 from the few vehicles and the city is highly diluted and dispersed pollutants (Kinney, et al., 2011) in  
8 regional air mass.

9

10 **Site 2: Tom Mboya Street, Fire station, Nairobi (urban roadside)**

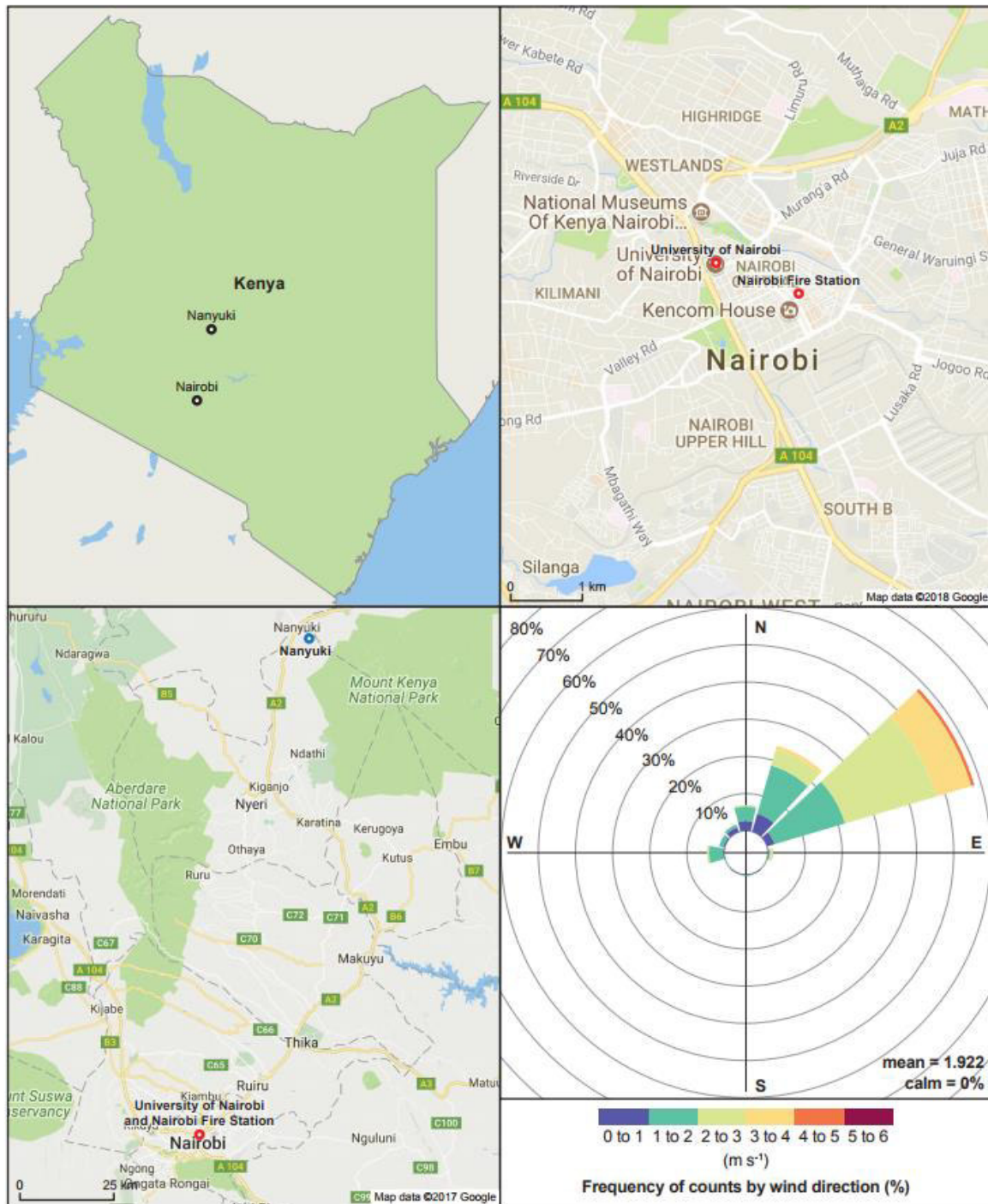
11 The second collection site in Nairobi was at the fire station, which is located within the CBD in the city.  
12 Unlike the American Wing site, the area around the Fire Station is characterized by high traffic flow  
13 which includes common public transport vans, locally known by the name "Matatus". It is on an urban  
14 street canyon, on a street where smoking diesel vans are frequent and is exposed to urban heat Island  
15 effects. It is also in the neighbourhood of vertical dispersion measurement site of PM<sub>2.5</sub> used by Kinney  
16 et al. (2011). The monitor was mounted at a height of approximately 5 m.

17

18 **Site 3: Nanyuki town (rural background)**

19 The third site chosen was on the outskirts of Nanyuki town, an administrative town in Laikipia County  
20 which is located to the North West of Mt. Kenya. The town is positioned at the Equator at  
21 approximately 1.28° S and 36.01° E. The highest point in Nanyuki is at an elevation of 2000 m above  
22 ground level. The town is home to approximately 50,000 people as per the last census conducted  
23 (KNBS, 2015). The Nanyuki region has a hinterland of significant agricultural cultivation, forest and  
24 considerable grazing activities (Gatari et al., 2005). The OPC was hung about 4 m above ground level  
25 thus exposing it to free regional air mass in an area of minimal local influence.





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**Figure 1. Locations of data collection sites and wind rose for the urban background location. Top left panel and bottom left panel: locations of Nairobi and Nanyuki in Kenya. Top right panel: shows relative locations of urban background location (University of Nairobi, American Wing) and urban roadside location (Nairobi fire station). Bottom right panel: wind rose generated from data collected at the urban background location during the measurement campaign.**

### 1 **3.2. PM Measurement Equipment**

2 Small low cost optical particle sensors (AlphaSense, OPC-N2, firmware version 18) were used to  
3 measure PM concentrations. The OPC-N2 is a miniaturized OPC which has dimensions of 75×60×65  
4 mm and weighs under 105 g. The unit cost of an OPC-N2 is approximately 250 GBP or 25000 KeS,  
5 hence it is significantly cheaper than reference optical particle counter instruments which cost  
6 approximately 30-50 times as much. Reference grade gravimetric instruments can cost even more.  
7 The lower cost of the OPC-N2, provided the opportunity for measurements at multiple sites  
8 simultaneously. It measures particles in the reported size range of 0.38 to 17 µm across 16 size bins,  
9 with a maximum particle count of 10,000 per second. The particle number concentration is converted  
10 by on-board factory calibration to PM concentrations according to European Standard EN481 (OPC-  
11 N2 manual).

12  
13 The assumed density for all particle sizes is 1.65 g/cm<sup>3</sup> and no special weighting is placed on any  
14 particular bin size. However, the manual for the OPC states “an additional weighting is applied on units  
15 with Firmware 18 or higher to account for under counting at low particle sizes and the effect of carbon  
16 particles in urban air so that the output matches collocated reference detectors.”

17  
18 The lower cut off for particle size observed by the OPC is 380 nm and hence a large proportion of all  
19 particles are not observed by the OPC due to the particle number being dominated by the smallest  
20 particle sizes (Seinfeld and Sypyros, 2016). Ultrafine particles (particles of aerodynamic diameter <100  
21 nm) were therefore not measured. However, the interest of the study was particulate mass which is  
22 dominated by particle sizes that were measured.

23  
24 The sensors had their data logged using Raspberry Pi 3 minicomputers. The Python script used to run  
25 the OPC-N2 on the Raspberry Pi 3 is discussed and provided in Crilley et al. (2018) and makes use of  
26 the py-opc python library for operating the OPC-N2 written by Hagan (2017). Together, the OPC-N2,  
27 minicomputer and accompanying wires and tubing were placed in bespoke weather resistant housing  
28 (dimensions ca. 30\*20\*10 cm). Power for both the OPC-N2 and minicomputer were provided by  
29 mains power.

30  
31 The OPC-N2 sensors are factory calibrated to measure PM mass concentrations representative of the  
32 UK. However, in our previous study, (Crilley et al., 2018) we demonstrated that in situ calibration of  
33 the sensors is required for the correct measurement of PM mass concentrations at urban background

1 sites in Birmingham, UK. The calibration in the Crilley et al. (2018) study involved both scaling and a  
2 relative humidity (RH) dependent term for when the RH is greater than approximately 85%.

3

4 The mass concentrations from the OPC-N2 devices, in the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size bins were recorded  
5 in time intervals of 10 s. For the subsequent analysis, the mass concentration data were aggregated  
6 into 1 h time-bins using the mean average. In time periods which contained missing data, the mean  
7 average of the available data was aggregated. All data manipulations were performed using R (version  
8 3.4.1), and the openair project package for R was used extensively for data visualization (Carslaw and  
9 Ropkins, 2012).

10

### 11 **3.3. Meteorological station**

12 The local meteorology for Nairobi was measured at the same location as the urban background site  
13 using a Vaisala instrument (WXT510) with the following variables measured: wind speed, wind  
14 direction, temperature, relative humidity, relative humidity, barometric pressure, and rainfall with an  
15 instrument temporal resolution of five minutes. The meteorology measured parameters were in good  
16 agreement with other local measurements such as those observed at Jomo Kenyatta International  
17 Airport (JKIA), which is approximately at an aerial distance of 10 km. The proximity of the  
18 meteorological station at the urban background site to the urban roadside makes the meteorological  
19 data appropriate for both sites. The data was collected at the urban background site from the 2<sup>nd</sup> of  
20 February to the 6<sup>th</sup> of April 2017, covering the duration of the PM measurements.

### 21 **3.4. OPC-N2 gravimetric mass calibration**

22 The OPC-N2 mass concentrations were calibrated using gravimetric measurements of PM<sub>2.5</sub> and PM<sub>10</sub>.  
23 The gravimetric calibration measurement was carried out on the 9<sup>th</sup> February 2017 for 24 h. A  
24 collocation measurement of the OPC and an Anderson dichotomous impactor (Sierra Instruments Inc.,  
25 USA) was set up, on the only possible date, at the background site. The impactor collected PM<sub>2.5</sub> and  
26 PM<sub>10-2.5</sub> particles on Teflon filters (diameter = 37 mm, pore size = 2 µm) at a total flow rate of 1 m<sup>3</sup> h<sup>-1</sup>.  
27 PM<sub>10</sub> is therefore the sum of the two size fractions (PM<sub>2.5</sub> + PM<sub>10-2.5</sub>). The chosen sample day was  
28 rain free and had similar temperature and RH profiles compared to the rest of the OPC sampling  
29 campaign. The filters were weighed using a mass balance before and after particulate matter  
30 collection. The observed 24 h average mass concentrations of PM<sub>2.5</sub> and PM<sub>10</sub> from the impactor were  
31 27.6 ± 6.8 and 51.8 ± 10.3 µg m<sup>-3</sup>, respectively, while those recorded from the OPC 16.9 and 30.6 µg  
32 m<sup>-3</sup>, respectively. The uncertainty in gravimetric concentrations was estimated from the instrument  
33 (10%), sampling (7%) and weighing (25%) errors and that of the OPC data was the standard deviation.

1 The value is as provided by the OPC, it is not possible to provide an error from the 1 day calibration.  
2 From the Crilley et al. (2018), the coefficient of variance is estimated as  $0.32 \pm 0.16$ ,  $0.25 \pm 0.14$  and  
3  $0.22 \pm 0.13$  for  $PM_{10}$ ,  $PM_{2.5}$  and  $PM_{10}$  mass concentrations, respectively. Hence, the observed scaling  
4 factors between the OPC derived masses and gravimetric analysis were 1.70 and 1.63 for  $PM_{10}$  and  
5  $PM_{2.5}$ , respectively. These factors are different to that observed in Crilley et al. (2018) which  
6 performed a similar gravimetric calibration procedure with the OPC-N2 measuring PM at an urban  
7 roadside sites in the UK. The discrepancies in scaling factors are likely due to differences in average  
8 particle densities observed in Kenya compared to that observed in the UK, and also the typical RH  
9 measured in Nairobi compared to the UK measurements (see discussion in next section). In particular,  
10 Nairobi PM has been shown to have a high percentage of mineral dust which typically has a high  
11 density, with Gaita et al. (2014) showing the annual average composition of  $PM_{2.5}$  being composed of  
12 35% mineral dust which originates from unpaved roads and wind-blown dust during the dry seasons.  
13 The gravimetric analysis did not allow for the calibration of the  $PM_{10}$  mass concentrations because a  
14 filter sample was not generated for the fraction of PM in this size range. Hence, the  $PM_{10}$  size fraction  
15 calibration uses the same calibration factor derived for the  $PM_{2.5}$  size fraction.

16  
17 The gravimetric calibration was carried out at the urban background field location, for the three OPC-  
18 N2s which were subsequently used in the measurement campaign at the three field sites. Hence, the  
19 calibration was most appropriate for the urban background site. Whilst the urban roadside site is in  
20 close proximity to the urban background site, the roadside site is more influenced by traffic related  
21 PM, hence, the average particle density at the roadside site is likely different to the urban background  
22 site. Likewise, the rural background site is likely to be far more influenced by mineral dust than the  
23 two urban sites. Hence the gravimetric calibration at the urban background sites only provides an  
24 estimate calibration for the urban roadside and rural background sites.

25  
26 Only one gravimetric calibration was carried out during the study period due to the lack of resource  
27 for further calibrations. If the PM composition varied significantly over the study period, then the true  
28 calibration factor will also change. Hence, the calibration factor used should be treated as an estimate  
29 for the whole study period because changes in PM composition lead to changes in particle refractive  
30 index, and therefore, the scattering pattern which is measured by the OPC to estimate particle size.  
31 Changes in particle density, due to compositional changes, also affects the particle mass calculated  
32 from the particle size. It is noted, for future studies it would be beneficial to have multiple gravimetric  
33 calibration points to check for continuing accuracy of the OPC-N2 sensors throughout the campaign.

34

### 1 **3.5. Measured particle mass dependence on relative humidity**

2 As detailed in Crilley et al. (2018), under UK conditions, the OPC-N2 device is sensitive to variations in  
3 RH when the RH exceeds ca. 85%. Crilley et al. (2018) suggest the RH dependence is due to the  
4 hygroscopic properties of particles that result in significant water mass being taken up by PM at high  
5 RH. This hygroscopic dependence can be modelled using a calibration that uses the  $\kappa$ -Kohler  
6 parameterization of aerosol hygroscopicity (Petters and Kreidenweis, 2008). The average  $\kappa$  parameter  
7 values for Africa ( $\kappa = 0.15 \pm 0.12$ ) are lower than for Europe ( $\kappa = 0.36 \pm 0.16$ ), as based on the Pringle et  
8 al. 2010 model, which is in good agreement with observational data. It is noted that composition of  
9 urban PM will have different hygroscopic properties to the average rural background. However, PM  
10 derived from urban emissions are often less hygroscopic than rural PM; therefore, the rural estimates  
11 might provide a useful upper estimate of particle hygroscopicity in urban centres. All locations used  
12 in the study period typically have RH less than the 85% threshold. However, it is noted that the RH  
13 dependent measurements shown in Crilley et al. (2018) were performed in the UK whereas these  
14 measurements were performed in Kenya. There may be significant differences between aerosol  
15 compositions, and hence hygroscopicities, in these two countries albeit both urban areas (Birmingham  
16 and Nairobi) will have significant vehicular influence. Measurements of RH at the Kenyan urban  
17 background site show that RH was only equal to or greater than 85% less than 1% of the time.  
18 Furthermore, there is no significant dependence of either the observed  $PM_{2.5}$  or  $PM_{10}$  mass  
19 concentration upon RH (see supplementary figures 1a and 1b), this is consistent with low  
20 hygroscopicity aerosols. The measurement period of work reported in this paper was in the Kenyan  
21 dry season with very few rain events, it is noted that if low cost sensors are to be used in the wet  
22 season in Kenya then the RH will likely be greater than 85% during significant periods and the  
23 hygroscopicity effect will likely need to be accounted for to obtain good measurements.

## 24 **4. Results**

### 25 **4.1. Particulate matter measurement**

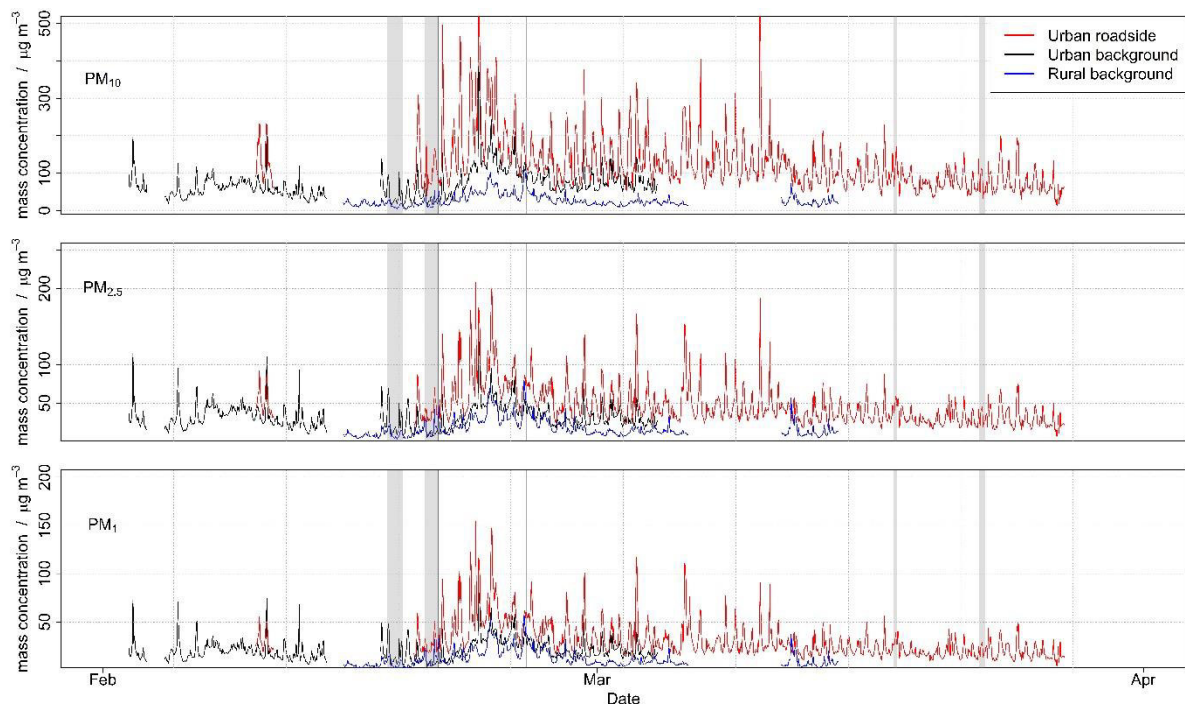
26 PM data was collected at the three sampling sites over the time period inclusive of 02/02/2017 and  
27 24/03/2017. Figure 2 provides the time series data for the  $PM_{10}$ ,  $PM_{2.5}$  and  $PM_1$  data over the whole  
28 measurement campaign. Gaps in data at specific sites are either due to the collocation of two or all  
29 three instruments at one site for cross calibration purposes, due to power failure requiring instrument  
30 restart or OPC malfunctioning.

31

32 The inter OPC-N2 precision was measured once during the campaign by co-locating the three  
33 instruments at the urban background site for 3 days for side by side sampling. The three instrument

1 colocation was carried out during at the start of the campaign (16/02/2017 – 18/02/2017). Two OPCs  
 2 were collocated together at the urban roadside site near the end of the campaign (04/03/2017 –  
 3 27/03/2017). All instruments gave very similar readings during both co-location periods, the inter-  
 4 instrument precision gave a coefficient of variance of  $8.8\pm 2.0\%$  in the  $PM_{2.5}$  fraction, with no  
 5 degradation in inter instrument precision observed over the sampling period. This coefficient of  
 6 variance is better than observed in Crilley et al. (2018) but this is expected because of the lower RH  
 7 conditions in Nairobi (see later discussion).

8



9

10 **Figure 2** Hourly time series data showing  $PM_{10}$ ,  $PM_{2.5}$  and  $PM_1$  mass concentrations at the three study  
 11 locations. Red line = urban roadside, black line = urban background and blue line = rural background.  
 12 Where multiple OPC-N2 devices were measuring in the same location at the same time, the average  
 13 is provided. The grey shading represents rain events as measured at the urban background location.

14 Continuous monitoring at all three sites was achieved for a fortnight in the period 18/02/2017 to  
 15 04/03/2017. This period will henceforth be referred to as the intensive period, whereas, the total  
 16 measurement campaign will be referred to as the campaign period. The number of monitoring days  
 17 for the urban roadside, urban background and rural background monitoring sites during the campaign  
 18 period were 40, 29 and 25 days, respectively.

19

20 Table 1 provides the average  $PM_1$ ,  $PM_{2.5}$  and  $PM_{10}$  mass concentrations observed at the three sites  
 21 during the campaign period. An identical table for the intensive period is included in the



1 supplementary material, see Table S1. The percentage of daily exceedances of daily PM<sub>2.5</sub> and PM<sub>10</sub> as  
 2 per WHO guidelines are also provided, however, to date there is no set guidelines of PM<sub>1</sub>. All  
 3 measurement sites exceeded the WHO daily guidelines for both PM<sub>2.5</sub> and PM<sub>10</sub> for some of the days  
 4 sampled. The urban roadside site exceeds the WHO guidelines on most days (85% for PM<sub>2.5</sub> and 90%  
 5 for PM<sub>10</sub>). Furthermore, on many days (13% for PM<sub>2.5</sub> and 40% for PM<sub>10</sub>) the urban roadside site  
 6 exceeds the WHO guidelines by at least twice as much. The urban background site has fewer  
 7 exceedances, compared to the urban roadside site, with daily exceedances occurring approximately  
 8 one third of the time. The urban background site is at an elevated position, which largely removes the  
 9 direct influence of local sources of PM pollution. As such, it can be assumed that the PM mass  
 10 concentrations observed at this location represent a lower limit for the ground level PM  
 11 concentrations throughout Nairobi, since most PM emissions will be due to ground level sources such  
 12 as vehicle emissions, fires, local industry and others. The rural background site has no daily  
 13 exceedances in the PM<sub>10</sub> size fraction but exceeds the PM<sub>2.5</sub> guidelines 12% of the time.

14

15 During the two-week intensive campaign, there was a period of elevated PM mass concentration  
 16 observed in PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size fractions centred around the 23<sup>rd</sup> February. The elevated PM  
 17 was observed in all three sites; therefore, **this might represent** a long-range pollution event.  
 18 Correspondingly, the average PM mass concentrations and percentage of WHO exceedances are  
 19 higher during the intensive period compared to the whole measurement campaign, see Table S1.

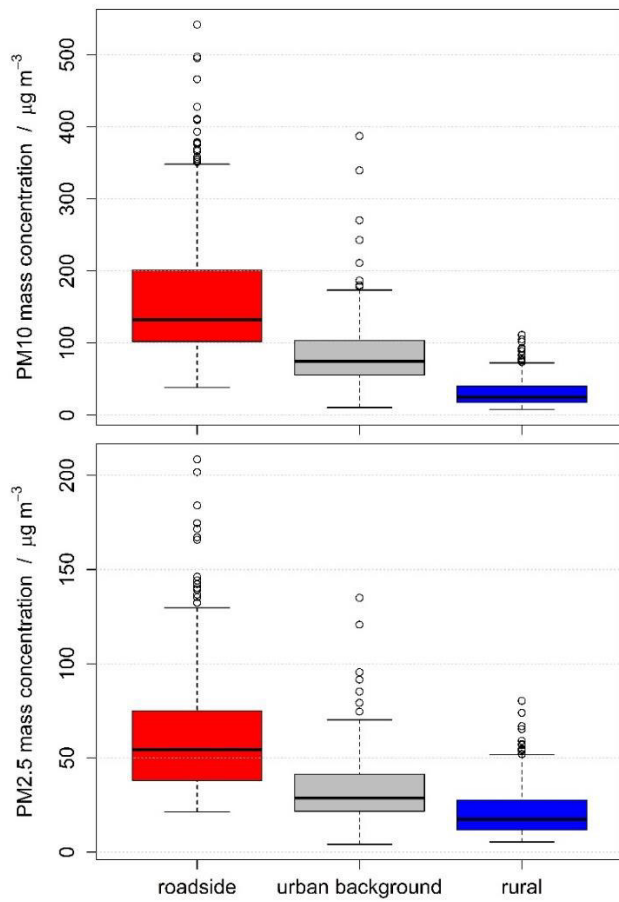
20 **Table 1** Mean average PM mass concentrations (PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub>) and daily exceedances of the  
 21 WHO PM guidelines (PM<sub>2.5</sub> and PM<sub>10</sub>) observed at the three measurement sites during the campaign  
 22 period. <sup>1</sup>WHO guidelines for daily PM<sub>10</sub> and PM<sub>2.5</sub> are 50 and 25 µg/m<sup>3</sup>, respectively

Measureme nt location	Measureme nt days (number)	Average PM <sub>1</sub> mass concentrati on (µg/m <sup>3</sup> )	Average PM <sub>2.5</sub> mass concentrati on (µg/m <sup>3</sup> )	Average PM <sub>10</sub> mass concentrati on (µg/m <sup>3</sup> )	% daily PM <sub>2.5</sub> exceedanc es <sup>1</sup>	% daily PM <sub>10</sub> exceedanc es <sup>1</sup>
Urban background	29	16.1	24.8	53.0	31.6	39.5
Urban roadside	40	23.9	36.6	93.7	85.0	90.0
Rural background	25	8.8	13.0	19.5	12.0	0.0

23

1 Whilst there is insufficient temporal data to provide a yearly average value for PM<sub>2.5</sub> and PM<sub>10</sub> mass  
2 concentrations for the three sites, the annual average values can be estimated from the data set using  
3 the average values provided in Table 1. These values are likely to be upper estimates for the yearly  
4 values because the measurements were obtained in period with little precipitation, thereby  
5 minimizing the degree of wet deposition of the PM. For instance, Gaita et al. (2014) showed that  
6 Nairobi's short rainy season (typically October – December) suppresses PM concentrations at the  
7 urban background site by approximately 50%. Notwithstanding the seasonal rain consideration, the  
8 average PM mass concentration observed in this study suggests that that the WHO recommendations  
9 for annual PM<sub>2.5</sub> and PM<sub>10</sub> are likely exceeded at both the urban background and urban roadside  
10 locations. For the urban background site, the measured average PM<sub>2.5</sub> and PM<sub>10</sub> mass concentrations  
11 exceed the annual WHO recommendations by factors of 2.5 and 2.7, respectively. Whereas for the  
12 urban roadside site they exceed recommendations by 3.7 and 4.7, respectively. These significant  
13 exceedances for both the urban roadside and urban background sites suggests that most of Nairobi's  
14 population will be subjected to outdoor air pollution far in excess of the WHO recommendations for  
15 annual exposure. Figure 3 provides the box and whisker plots for the hourly averaged PM<sub>2.5</sub> and PM<sub>10</sub>  
16 data for the three measurement sites. Figure S3 provides box and whisker plots for the daily averaged  
17 PM<sub>2.5</sub> and PM<sub>10</sub> data for the three measurement sites and highlights the proportion of the days which  
18 exceed the WHO annual and daily recommendations.

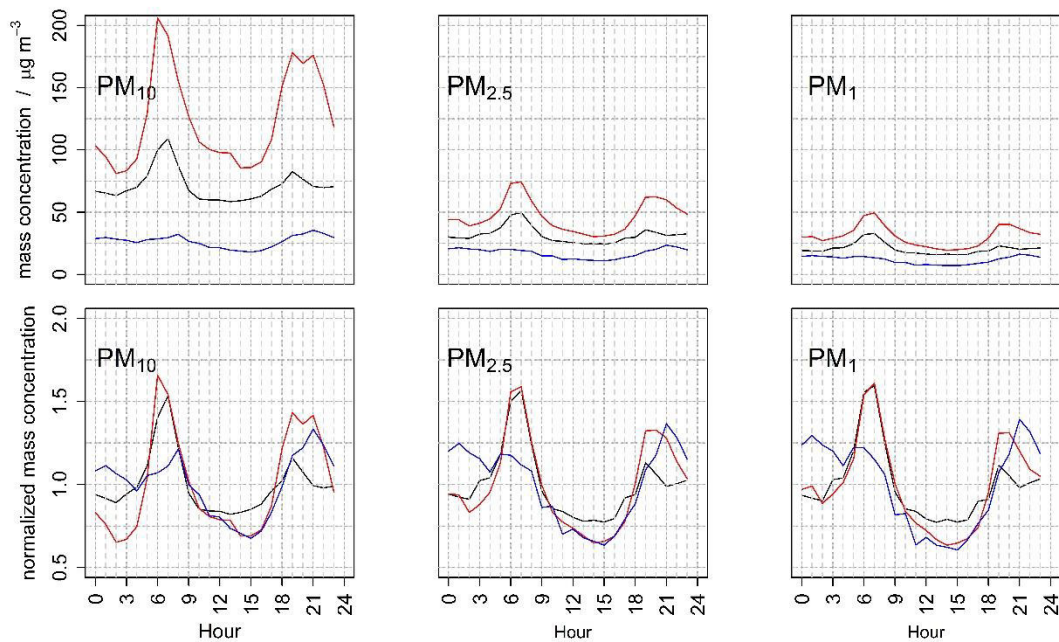




1

2 **Figure 3** Box and whisker plots of the hourly averaged PM<sub>2.5</sub> and PM<sub>10</sub> mass concentrations measured  
 3 at the three sites. The green dashed and dotted lines represents the WHO recommended annual and  
 4 daily limits, respectively

5



1

2 **Figure 4** Diurnal variation in PM<sub>10</sub>, PM<sub>2.5</sub> and PM<sub>1</sub> mass concentration measured at the three field sites  
 3 during the whole campaign period. Top panels show the measured concentrations. Bottom panels  
 4 shows the mass concentrations that have been normalized to the average mass concentration.

5

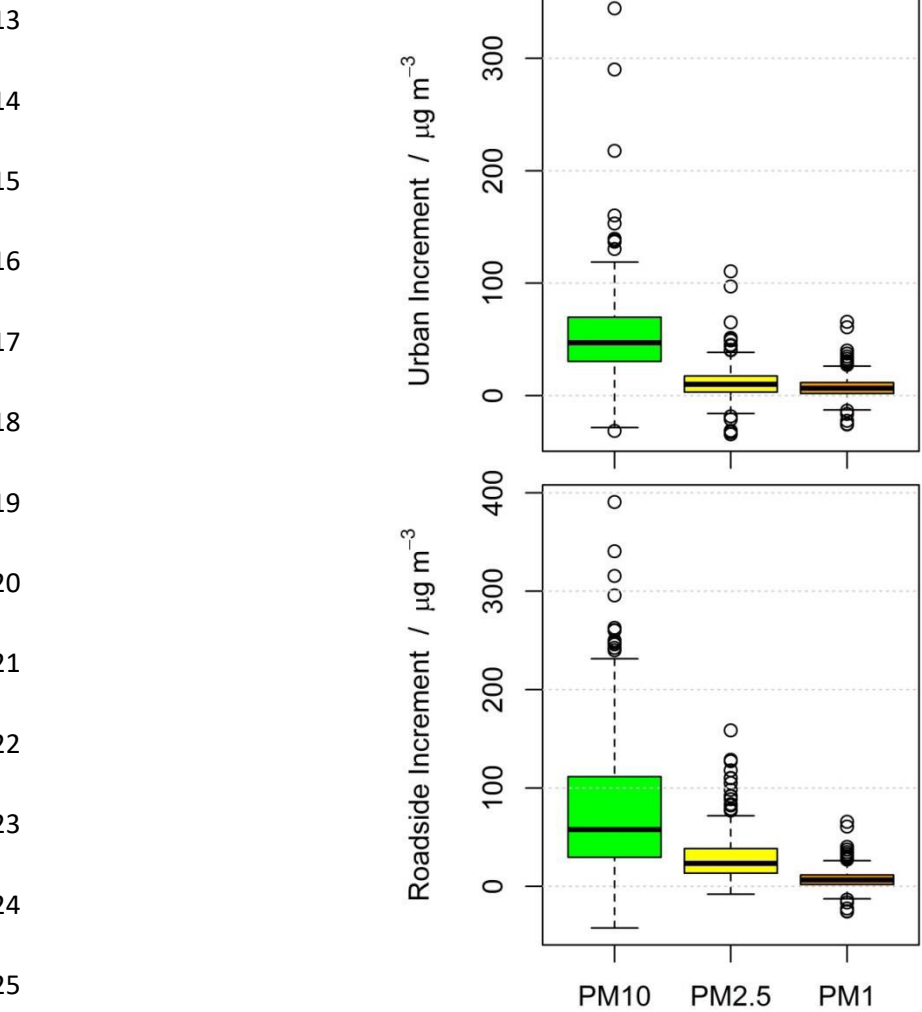
6 Figure 4 provides the mean average diurnal hourly profiles of the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> mass  
 7 concentrations for the three measurement sites during the whole campaign period. There is clear  
 8 diurnal variation observed at all the sites, two distinct peaks are observed in the two urban locations  
 9 during the morning (ca. 05:00 – 10:00) and the evening (ca. 18:00 – 24:00) which correspond to the  
 10 Nairobi peak traffic periods. The normalized data shows that the traffic related structure is very similar  
 11 in both the urban background and urban roadside sites indicating that the traffic related PM pollution  
 12 is the dominant source at both sites. The rural background site also shows diurnal variation with some  
 13 indication of a traffic related signal at similar times to the urban sites, especially in the PM<sub>2.5</sub> size  
 14 fraction. However, overall the rural diurnal cycle appears to largely correspond to solar insolation  
 15 suggesting the dominant factor affecting the rural mass concentrations is the height of the local  
 16 boundary layer which decreases in the night time and increases with greater solar insolation.

17

18 Through comparison of the urban roadside, urban background and rural background hourly averaged  
 19 data, it is possible to generate estimates of urban increments and roadside increments relevant for  
 20 Nairobi using a 'Lenschow' type approach (Lenschow et al., 2001). For the intensive period the urban

1 and rural increments are calculated for both the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> mass concentrations, see Figure  
2 5. The urban increment is calculated by subtracting the hourly average values of the rural background  
3 site from the urban background site. During the intensive period, analysis of the air mass back  
4 trajectories indicates that the regional wind direction was almost exclusively from the northeast.  
5 Hence the Nanyuki rural background site is a good representative of the rural background that impacts  
6 upon Nairobi.

7 The roadside increment was calculated by subtracting the hourly average values of the urban  
8 background site from the urban roadside site. It is noted that the chosen roadside measurement site  
9 is particularly busy with vehicles, compared to many other non-highway streets in Nairobi. In  
10 particular, the site is a popular Matatu (14 seat passenger vans) terminal with multiple vehicles idling  
11 at any point during the day. Therefore, the roadside increment obtained using this location likely  
12 represents a value close to the upper boundary for Nairobi roads.



26 **Figure 5** Box and whisker plots of urban and roadside increment of PM<sub>10</sub>, PM<sub>2.5</sub> and PM<sub>1</sub> calculated for  
27 Nairobi. Data was taken from the intensive campaign period when the urban background, urban

1 roadside and rural background sites were all measuring simultaneously. Hourly averaged mass  
2 concentration data is used

3 The urban and roadside increments are significant for all the investigated PM size fractions. A  
4 statistical summary of the roadside and urban increments for the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size fractions  
5 are given in Table 2.

6 **Table 2** Summary of roadside and urban increments for the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size fractions  
7 measured during the intensive period.

	Roadside Increment ( $\mu\text{g}/\text{m}^3$ )			Urban Increment ( $\mu\text{g}/\text{m}^3$ )		
	PM <sub>1</sub>	PM <sub>2.5</sub>	PM <sub>10</sub>	PM <sub>1</sub>	PM <sub>2.5</sub>	PM <sub>10</sub>
Minimum	-4.3	-6.1	-31.7	-20.1	-26.9	-23.6
1 <sup>st</sup> Quartile	7.3	10.5	22.2	1.0	2.2	19.5
Median	12.6	18.3	43.3	4.7	7.1	33.1
Mean	18.9	22.9	58.1	5.2	8.2	36.6
3 <sup>rd</sup> Quartile	20.7	30.0	83.4	8.7	13.2	48.2
Maximum	95.5	123.9	292.6	51.2	86.3	258.0

8  
9 During the intensive period, the mean average roadside increment is 57.2, 47.5 and 48.1 % of the  
10 mean roadside mass concentration, in the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size fractions, respectively.

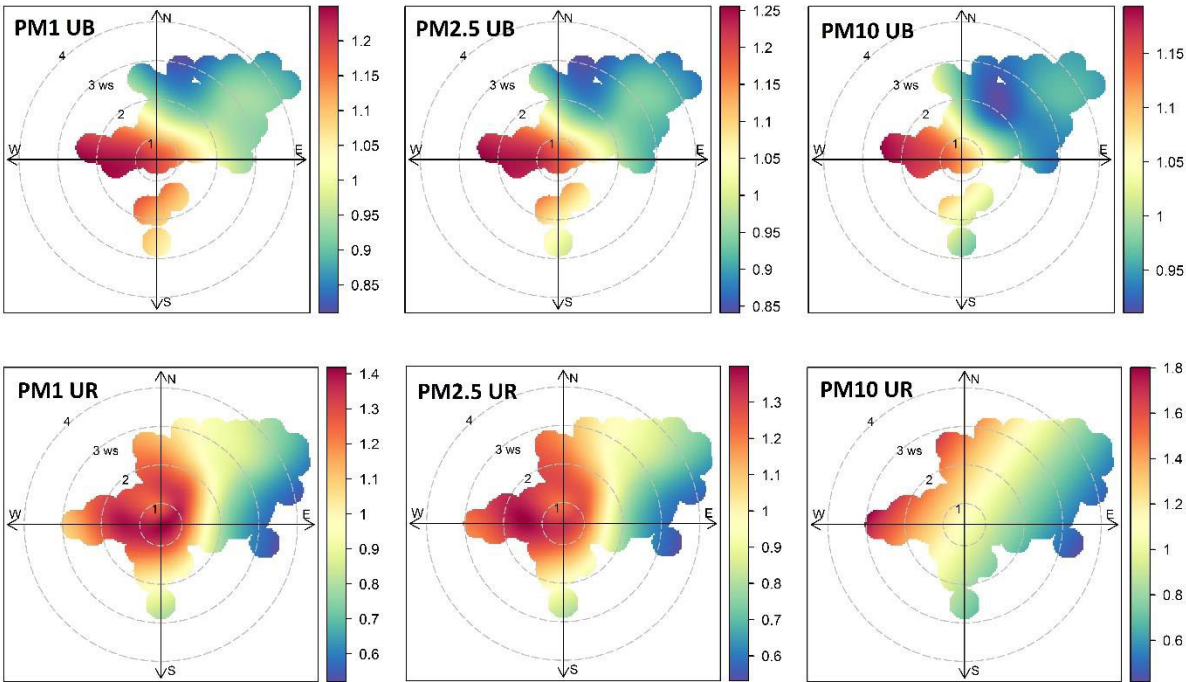
11 The spatial variation in PM emissions, in the different size fractions, can be assessed at the urban  
12 background and urban roadside sites using bivariate polar plots, which provide information on the  
13 variation of PM mass concentration with wind direction and speed, see Figure 6 (Carslaw and Beevers,  
14 2013). The urban background and urban roadside sites are sufficiently closely collocated (< 0.5 km  
15 apart) that the wind data acquired at the urban background site is applicable to the urban roadside  
16 site. Wind direction data was not available for the rural background site, so analysis of the spatial  
17 variation was impossible at this site.

18 Figure 6 clearly shows significant variation of PM mass concentration at both urban sites, which are  
19 dependent upon the wind conditions. The urban background site shows broadly similar behaviour in  
20 the spatial variation of the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> size fractions. The peak in concentrations are observed  
21 at low wind speeds and when the wind comes from the west and south. This wind direction  
22 dependence is consistent with the close proximity of the major highway A104 'Nairobi-Malaba Road',  
23 which passes close to the site in the direction of high PM concentrations. The diurnal profiles and  
24 roadside increments discussed earlier combined with the wind dependence highlights the role of

1 roads in Nairobi as the major source of PM in all size fractions studied. Since the site is within Nairobi's  
2 Central Business District (CBD), there are other significant roads nearby as well, but the A104 has the  
3 greatest fleet density.

4 The urban roadside site also shows distinct variation in pollutant concentrations with wind speed and  
5 direction. In the  $PM_{10}$  size fraction the greatest concentrations are seen to the northwest and smallest  
6 to the southwest with a steady reduction between these two extremes. The  $PM_{2.5}$  and  $PM_1$  size  
7 fractions show a more complex behaviour with highest concentrations at low wind speeds and the  
8 north and west directions. The urban roadside location is surrounded by small roads and lower traffic  
9 speeds compared to the highways, for example the A104. The lower traffic speeds likely lead to less  
10 non-tail pipe emissions from dust resuspension and hence there are less local  $PM_{10}$  particles when  
11 compared to the urban background site. Whereas the localized  $PM_{2.5}$  and  $PM_1$  concentrations are  
12 likely due to the heavily congested local roads on which Matatus and other vehicles are often left  
13 idling leading to high tail pipe emissions, which are typically in the smaller PM size fractions (Pant and  
14 Harrison, 2013).

15

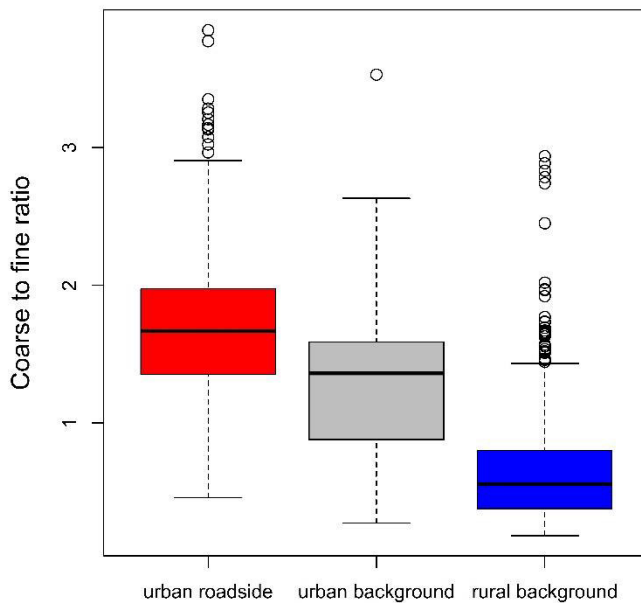


16

17 **Figure 6** Bivariate (polar) plots of PM in different size fractions at urban background (UB) and urban  
18 roadside (UR) sites. Panel descriptions A)  $PM_1$  UB, B)  $PM_{2.5}$  UB, C)  $PM_{10}$  UB, D)  $PM_1$  UR, E)  $PM_{2.5}$  UB and  
19 F)  $PM_{10}$  UR. The PM mass concentration data in each plot are normalized to allow for easy comparison  
20 between the different sites and PM size fractions investigated. However, note the scale bars are  
21 different for each panel to allow for easier interpretation.

1

2 Figure 7 provides the distributions of the ratio between the coarse and the fine PM mass fractions at  
3 the three field sites during the campaign period. Whilst each site shows distinct variation, with large  
4 interquartile ranges, in the reported coarse:fine ratio, which is dependent on the time of year and  
5 time of day, the median ratios at each site are distinct, with the ratio at the urban roadside, urban  
6 background and rural background sites being 1.6, 1.3, and 0.5, respectively. At the roadside site, the  
7 median coarse:fine mass ratio is almost triple that observed at the rural background; this is consistent  
8 with the dominant source of PM at the roadside site being the **resuspension of large dust particles**. At  
9 the rural site, the PM size distribution has a greater ratio of fine material consistent with the rural site  
10 having a signature of regional background PM. The ratio of coarse:fine PM at the urban background  
11 site is intermediate between the roadside and rural background sites which suggests that this site is  
12 effected significantly by both the regional background and the urban road PM sources. These insights  
13 into the coarse:fine PM ratio is consistent with the roadside and urban increments, shown in Figure 5  
14 and discussed previously.



15

16 **However, PM derived**

### 17 **4.3. Comparison with previous measurements**

18 To the best of our knowledge, there has been no previous literature study to date utilising calibrated  
19 low cost sensors to measure PM in Nairobi. Furthermore, it is difficult to make comparison with

1 previous Nairobi based PM studies because of the differences in the temporal resolution of the data  
2 and campaign durations used in this study compared to past measurements.

3 The most comparable study of PM<sub>2.5</sub> would be the work of Gaita et al., (2016) which also recorded the  
4 levels of PM<sub>2.5</sub> at the University of Nairobi (urban background site). The urban background average of  
5 PM<sub>2.5</sub> during this study's campaign period was 24.8 µg/m<sup>3</sup> compared to Gaita et al., (2016) mean  
6 average of 9.8µg/m<sup>3</sup>, showing a significant increase of 253%. The sampling time window used in Gaita  
7 et al., (2016) study was between August to September 2007 which is distinct from the February to  
8 March 2017 period of this study. Both of these study periods were largely dry, with low precipitation  
9 levels, thereby suggesting PM deposition would have been similar between the two studies. The  
10 significant increase in measured PM<sub>2.5</sub> could be due to several reasons. Firstly, there could be seasonal  
11 differences between August/September and the February/March sampling periods of the two studies;  
12 however, the study of Gaita et al. (2014) suggests the urban background concentrations of PM<sub>2.5</sub> mass  
13 concentration is similar between these two time periods. The regional background PM loading may  
14 have increased during this time period, potentially due to increasing regional aridity caused by climate  
15 change leading to more dust generation (Greve et al., 2017). There is almost ten years difference in  
16 the times of this study compared to Gaita et al. 2016, in this time Nairobi has undergone significant  
17 increases in population and urbanization with correspondingly higher use of motorization and fuel.  
18 Using UN population data (UN, 2014), the population of Nairobi is well modelled by equation E1, in  
19 which  $Y$  is year date, and  $p$  is the population in thousands, which suggests that the population of  
20 Nairobi has increased by 148% from 2007 to 2017.

$$21 \quad p = 2.33 \times 10^{31} \exp(3.91 \times 10^{-2} \times Y) \quad (E1)$$

22 Hence, the population increase alone cannot account for the increase in PM concentration. The  
23 pollution production capability per capita could have increased, which is very likely because of the  
24 increased rates of motorization and fuel use. If we assume that the increase in PM is solely due to  
25 population increase and per capita pollution, it suggests that in 2017 the average citizen is 70% more  
26 polluting than the average citizen in 2007.

27 The Egondi et al. (2016) study of PM<sub>2.5</sub> in two slums in Nairobi reported much higher values of 166  
28 µg/m<sup>3</sup> and 67 µg/m<sup>3</sup> for two different slum areas within Nairobi. These values are much higher than  
29 the average PM<sub>2.5</sub> values from this study; Egondi et al. stated that the reason for such high levels of  
30 PM<sub>2.5</sub> stemmed from the local situation and distinct sources of PM within the two slums. This study  
31 used a TSI optical particle counter, which was placed 1.5 m above ground level. Therefore, these  
32 measurements were likely highly influenced by re-suspended dust.

1 Although Kinney et al., (2012) measured PM<sub>2.5</sub> levels at four roadside locations, the sampling window  
2 was only 11 hours and therefore it is not possible to directly compare this study to it. However,  
3 considering the diurnal variation in PM found in this study, both investigations measured similar PM<sub>2.5</sub>  
4 levels. Kinney et al., (2012) recorded daytime concentration ranges of 10.7 µg/m<sup>3</sup> and 98.1 µg/m<sup>3</sup> for  
5 a rural and urban roadside site, respectively, compared to ca. 25 µg/m<sup>3</sup> and ca. 150 µg/m<sup>3</sup> for this  
6 study. Again, the increase between sampling years may be a reflection of the increased population,  
7 vehicular traffic and rapid urbanisation.

## 8 **5. Discussion**

9 In this study, we have shown that Nairobi currently has very high levels of PM mass concentration in  
10 the PM<sub>1</sub>, PM<sub>2.5</sub> and PM<sub>10</sub> mass fractions. These measurements were conducted using low cost  
11 calibrated OPC-N2 sensors. The measured PM<sub>2.5</sub> and PM<sub>10</sub> concentrations at the urban roadside and  
12 urban background sites both regularly exceeded the WHO daily limits and very likely exceed the annual  
13 limits. In particular, the roadside site often showed concentrations of double the WHO guidelines.  
14 These concentrations will very likely be causing significant harm to the population of the Nairobi.

15 The negative health effect of PM is linked to the level of exposure experienced by the patient. This  
16 paper and others (e.g. Gaita et al. 2014) have shown that in Nairobi, vehicle emissions are the most  
17 significant source of PM. Hence, in Nairobi and other similar cities, the exposure to outdoor PM is to  
18 a large extent a function of ones proximity to roads. Furthermore, since traffic varies diurnally,  
19 seasonally and by day of the week, personal exposure is both spatially and temporally dependent. This  
20 spatial and temporal heterogeneity leads to health inequalities in cities. The urban poor who are often  
21 most vulnerable to environmental risks due to lack of adequate health provision, typically live in close  
22 proximity to roadways, heightening their exposure to vehicular emissions. Stemming from poorly  
23 planned rapid urbanisation and inadequate service provision within these cities, those that are unable  
24 to afford public transport or personal vehicles frequently walk along these pollution heavy roads, only  
25 increasing their exposure periods (Hajat et al., 2015).

26 This study only looked at outdoor air quality, it is important to stress that a significant amount of air  
27 pollution deaths in Kenya and SSA in general are due to poor indoor air quality (Mannucci and  
28 Franchini, 2017). As a total number of deaths, deaths related to indoor air quality in Kenya rose to  
29 18% from 1990-2013 (Roy, 2016). In LMIC countries, indoor exposure to pollutants is typically from  
30 the household combustion of solid fuels on open fires or traditional stoves (Yip et al., 2017). These  
31 exposures increase the risk of acute lower respiratory infections and associated mortality among



1 young children; indoor air pollution from solid fuel use is also a major risk factor for cardiovascular  
2 disease, chronic obstructive pulmonary disease and lung cancer among adults (Muindi et al., 2016).

3

4 This study has shown that the low cost OPC-N2 sensors can be used to generate diurnal PM datasets  
5 with good precision and repeatability. As noted in the methodology, it would have been preferable  
6 for more cross calibration periods with the tried and tested gravimetric PM measurement but  
7 resources did not allow this. In addition to more calibration points, the study could have been  
8 enhanced by the inclusion of collocated calibration points for the roadside and rural background sites  
9 in addition to the urban background site, since the average particle shape, size and density will likely  
10 be different between the three sites because of differing PM sources and emission factors. However,  
11 it is noted, that whilst it is desirable from a purely scientific point of view to have more inter-  
12 comparison with reference grade equipment; every inter-comparison adds significant additional cost  
13 to the project both in terms of consumables for the gravimetric analysis (including the cost of  
14 analytical grade filters and accompanying laboratory supplies), and the cost in manpower. Many other  
15 cities in SSA and other LMIC countries do not have the resource that Nairobi does in having a  
16 gravimetric sampler. These additional costs required for highly accurate scientific results would likely  
17 make the low cost sensors not so very low cost after all, and hence bring into question their unique  
18 selling point (USP).

19 Whilst this paper focused on PM pollution, it is noted, that there are serious risks to health not only  
20 from exposure to PM, but also from exposure to ozone (O<sub>3</sub>), nitrogen dioxide (NO<sub>2</sub>) and sulfur dioxide  
21 (SO<sub>2</sub>) (Lelieveld et al., 2015). As with PM, concentrations are often highest largely in the urban areas  
22 of low- and middle-income countries. Ozone is a major factor in asthma morbidity and mortality, while  
23 nitrogen dioxide and sulfur dioxide also can play a role in asthma, bronchial symptoms, lung  
24 inflammation and reduced lung function. Good quality measurements of these gas phase pollutants  
25 lag behind measurements of PM in Nairobi, other SSA cities and LMIC cities in general. This is due to  
26 the high importance of PM as an environmental risk factor but also because of the lack of good quality  
27 gas analysers which are affordable and transportable.

## 28 **6. Conclusions**

29 Air quality in many LMIC urban centres is often poor and in many cities is getting worse due to the  
30 combined pressures of increasing population, increasing urbanization, increasing vehicular traffic and  
31 poor vehicle regulation. To be able to manage air pollution, good quality and long term data sets are  
32 required. Unfortunately, in many LMICs the cost of certified high quality air quality measurements is

1 beyond the financial means of environmental authorities. Low cost sensors offer the possibility of air  
2 quality products at significantly lower cost compared to traditional methods.

3 This paper used calibrated OPC-N2 devices to measure PM concentrations in Nairobi, Kenya in the size  
4 fractions  $PM_{1}$ ,  $PM_{2.5}$  and  $PM_{10}$ . The data required calibration using an established gravimetric  
5 approach to PM measurement. The need for calibration by trained scientists significantly increases  
6 the costs associated with low cost monitoring and this cost needs to be factored in when assessing  
7 options for air quality monitoring.

8 PM was measured in three locations: an urban roadside, urban background and rural background site  
9 for a period of approximately two months. The data reveals that roadside and urban background  
10 locations in Nairobi often exceed the WHO guidelines for daily averaged PM mass concentration in  
11 both the  $PM_{2.5}$  and  $PM_{10}$  size fractions. Comparison of the data with previous measurements  
12 conducted in Nairobi is difficult but where comparison is possible, it appears that air quality has  
13 become worse in the last ten years which is likely due to increases in population, urbanisation and  
14 motorization. **Changes in industry may also influence the air quality.** Comparison of the data from  
15 the three sites, following a 'Lenschow' type approach, allowed for the calculation of representative  
16 roadside and urban increments for Nairobi (Lenschow et al., 2001). This increment data can be used  
17 in future air quality modelling to assess the likely health impact of PM pollution on Nairobi's  
18 population. The combination of the diurnal PM data with local meteorology allows for simple source  
19 apportionment of the PM. The diurnal PM concentrations tracks the Nairobi rush hours, furthermore,  
20 PM peaks when the wind comes from the direction of significant numbers of vehicles such as major  
21 roads and a Matatu stop. These facts taken together, point towards vehicle emissions being the major  
22 sources of air pollution in Nairobi, as has been previously observed in studies such as Gaita et al. 2014.  
23 The coarse PM fraction increases at roadside compared to urban background site suggesting that non-  
24 exhaust vehicle emissions make up a significant amount of the vehicle emissions.

25 In summary, the low cost sensors used in this study provided much useful data for assessing air quality  
26 in Nairobi at an equipment cost significantly lower than that of traditional instruments. Low cost  
27 sensors have great potential in other country settings and could be used for long term sampling if the  
28 appropriate calibrations are performed.

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