Supplementary Information (SI) for Environmental Science: Atmospheres. This journal is © The Royal Society of Chemistry 2024

Supplementary Data

2.0 Supplementary Data: Method

2.4 Data Preprocessing

The raw data from AirQo and TAHMO for PM_{2.5} and meteorological variables were processed

and prepared for subsequent analysis. The data manipulation and pre-processing were

performed using Python version 3.10.11(73) and several libraries, including Pandas version

2.0.2, NumPy version 1.23.5, and Matplotlib version 3.7.0.

2.4.1 PM_{2.5} Data Preprocessing

The PM_{2.5} data obtained from AirQo for Jinja and Kampala were analysed using separate

Jupyter notebooks. Each dataset contained hourly timestamp information, PM_{2.5} raw values,

PM_{2.5} calibrated values, and site names. Jinja had seven sites where the AirQo sensors were

deployed, while Kampala had 55 unique sites. The detailed sites where the AirQo sensors were

deployed in the two cities are presented in Table 1.

Only data from January 2020 to December 2022 were considered to ensure consistency and

reliability. Data from 2019 were excluded due to limited availability and potential data

inconsistencies. The next step involved averaging the hourly PM_{2.5} data across all sites within

each city, resulting in city-level PM_{2.5} concentrations. This aggregation was performed because

the analysis focused on the overall PM_{2.5} levels in each city.

After this, the percentage of missing values was calculated for both Jinja and Kampala datasets.

The missing data statistics are recorded in Table 2. The Multiple Imputation by Chained

Equation (MICE) algorithm was utilized to impute the missing values. MICE is a multivariate

imputation method that takes into account the relationships among variables in the dataset. It

generates multiple plausible imputations for missing values by fitting regression models for the

incomplete variable using the remaining variables as predictors(74). The imputation process is

performed iteratively, resulting in a completed dataset with imputed values. MICE is

particularly useful in handling missing data and can handle missingness up to 30%(75, 76).

2.4.2 Meteorological Data Preprocessing

The raw meteorological data obtained from TAHMO for Jinja and Kampala were analysed using separate Jupyter notebooks. Each dataset contained hourly meteorological measurements of atmospheric pressure (kPA), precipitation (mm), temperature average (°C), wind gusts max (m/s), wind speed(m/s), and relative humidity. To ensure consistency with the PM2.5 data, only data from January 2020 to December 2022 were considered, excluding 2019. Some meteorological parameters, including radiation average (W/m), temperature min (°C), temperature max (degrees Celsius), and wind direction (°), were excluded from the datasets as they were deemed unnecessary for the analysis. The percentage of missing values was calculated to handle missing data in the meteorological datasets, similar to the PM_{2.5} data. The missing data statistics are presented in Table 2. The MICE algorithm was then employed to impute the missing values, ensuring a comprehensive dataset for subsequent analysis.

2.4.3 Integration of PM_{2.5} and Meteorological Data

After completing the individual pre-processing steps for both cities' $PM_{2.5}$ and meteorological data, the datasets were joined on the timestamps. The integrated dataset now consisted of $PM_{2.5}$ calibrated values and the various meteorological variables aligned in time, providing a comprehensive dataset for further analysis. The final dataset contained a matrix of 26,304 rows x 8 columns.

Supplementary File SF 1: Sites of AirQo devices in Kampala and Jinja.

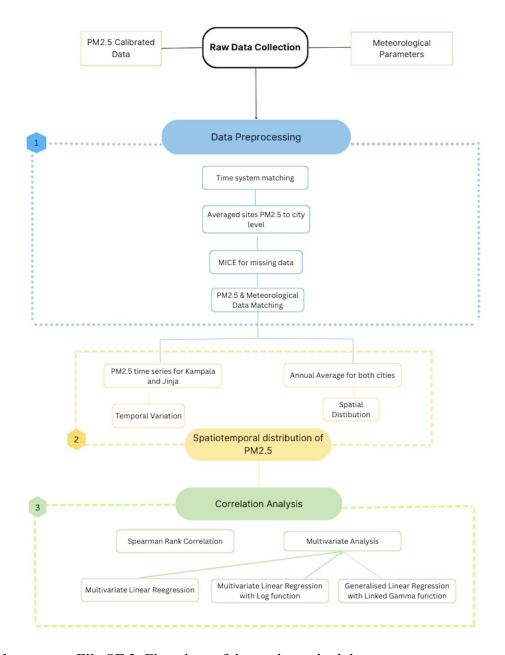
City	Sites
Jinja	Jinja Main Street, Kimaka, City Council, Walukuba East, Mpumudde, Rubaga Ward, YMCA
Kampala	Civic Centre Kampala Central', Bugolobi Nakawa, Rubaga Kampala, Kasubi Rubaga, Bukasa Kira Municipality, Bukoto I Kawempe, Nsambya/US Embassy Makindye, Kalerwe, Kawempe, Butabika Nakawa, Kiwatule Nakawa, Makerere University_01 Kawempe, Nakasero II Kampala, Mbuya II Nakawa, Banda Kampala, Luwafu Makindye, Ntinda Nakawa, Makerere University Weather Station, Bahai Kawempe, Kawempe Industrial Kawempe, Mpererwe

Kawempe, Kyanja Nakawa, Luzira Nakawa, Rubaga, Kampala Central, Makindye, Kyebando Kawempe, Kawempe, Nakawa, Kampala Central, Makerere, Kawempe, Lower Nsooba, Kawempe, Nakawa, Nakawa, Makindye Lukuli Makindye, Lukuli Makindye, Makindye Nsambya/US Embassy, Busega Rubaga, Makindye KCCA division offices, Kibuli Makindye, Kizungu, Makindye, Ggaba Makindye, St.Ponsiano primary school Makindye, wabigalo, Kansanga Seed Secondary School, Kampala Uganda, Muyenga tank hill ,makindye, Kabalagala town Makindye, Katwe 1, Makindye, Kayan Test Site , Naguru II Nakawa, Kisugu Makindye, Kyebando Kampala, Rubaga Div Rubaga, Katwe 1 Makindye, Muyenga tank hill, Makindye, City parents primary school

Supplementary File SF 2: Missing Data Statistics for PM_{2.5} and Meteorological Parameters in Kampala and Jinja

	Percentage of Missing Data for each Parameter (in %)						
Study	PM _{2.5}	Atmospheric	Precipitation	Relative	Average	Wind	Wind
city		Pressure		Humidity	Temperature	speed	gusts
Kampala	16.54	15.37	0.00	15.43	15.37	16.29	16.29
Jinja	24.86	12.59	23.15	12.61	12.57	19.31	19.31

Since the imputation is based on neighbouring values and a significant number of data points were missing in that period, the imputation resorted to using the mean value, resulting in a flat trend. To reduce the noise in the dataset and provide a clearer depiction of the trend, a 30-day moving average was calculated, as shown in Figure 5.



Supplementary File SF 3: Flowchart of the study methodology

Supplementary data: Complete Model Results

MULTIVARIATE LINEAR REGRESSION

Kampala

OLS Regression Results

Dep. Variable:	pm2_5_	calibrated_va	lue	R-sq	uared:		0.169
Model:			0LS	Adj.	R-squared:		0.169
Method:		Least Squa	res		atistic:		1339.
Date:		Sat, 15 Jul 2	023	Prob	(F-statistic):	0.00
Time:		00:52	:06	Log-	Likelihood:		-1.1217e+05
No. Observation	s:	26	304	AIC:			2.244e+05
Df Residuals:		26	299	BIC:			2.244e+05
Df Model:			4				
Covariance Type	:	nonrob	ust				
	coef	std err		+	D> +	[0.025	0.0751
	coei	sta err		t	P> t	[0.025	0.975]
const	208.6426	3.015	69.	206	0.000	202.733	214.552
precipitation	-0.5126	0.091	-5.	655	0.000	-0.690	-0.335
humidity	-86.8701	1.754	-49.	527	0.000	-90.308	-83.432
temp_mean	-3.9473	0.080	-49.	170	0.000	-4.105	-3.790
wind_speed		0.413	-39.	711	0.000	-17.203	-15.585
Omnibus:		6112.635	Dur	bin-W	======== atson:		0.233
Prob(Omnibus):		0.000			era (JB):	1	6854.350
Skew:		1.236		(JB)		_	0.00
Kurtosis:		6.044		d. No			724.
=======================================						=======	

Jinja

OLS Regression Results

Dep. Variable:	pm2_5_6	calibrated_va	alue R-s	quared:		0.06
Model:			OLS Adj	. R-squared:	0.06	
Method:		Least Squa	ares F-s	tatistic:		450.
Date:	:	Sat, 15 Jul 2		b (F-statist:	ic):	0.0
Time:		00:48	3:46 Log	-Likelihood:		-1.0556e+0
No. Observation	is:	26	304 AIC	:		2.111e+0
Df Residuals:		26	5299 BIC	:		2.112e+0
Df Model:			4			
Covariance Type	::	nonrob	oust			
	coef	std err	t	P> t	[0.025	0.975]
const	102.9843	2.187	47.081	0.000	98.697	107.272
precipitation	-0.7912	0.079	-10.056	0.000	-0.945	-0.637
humidity	-40.1345	1.376	-29.171	0.000	-42.831	-37.438
temp_mean	-2.1865	0.055	-40.064	0.000	-2.293	-2.080
wind_speed	0.7491	0.074	10.128	0.000	0.604	0.894
Omnibus:		16282.613	Durbin-			0.535
Prob(Omnibus):		0.000	Jarque-	Bera (JB):	29	4530.187
Skew:		2.668	Prob(JB			0.00
Kurtosis:		18.501	Cond. N	0.		710.

MULTIVARIATE LINEAR REGRESSION WITH LOG OF TARGET VARIABLE

Kampala

OLS Regression R	esults
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Dep. Variable:	pm2_5_	calibrated_va	lue	R-squ	ared:		0.17
Model:			0LS	Adj. R-squared:			0.17
Method:		Least Squa	res	F-sta	tistic:		1418
Date:		Sat, 15 Jul 2023		Prob	(F-statisti	c):	0.00
Time:		00:54	:12	Log-L	ikelihood:		-13267.
No. Observations	:	26	304	AIC:			2.654e+0
Df Residuals:		26	299	BIC:			2.658e+04
Df Model:			4				
Covariance Type:		nonrob	ust				
	coef	std err		t	P> t	[0.025	0.975]
const	7.4381	0.070	105.	978	0.000	7.301	7.576
precipitation	-0.0135	0.002	-6.	412	0.000	-0.018	-0.009
humidity	-1.9676	0.041	-48.	185	0.000	-2.048	-1.888
temp mean	-0.0891	0.002	-47.	680	0.000	-0.093	-0.085
wind_speed	-0.4199	0.010	-43.	691	0.000	-0.439	-0.401
======================================	======	19.786	Dur	bin-Wa	======= tson:		0.230
Prob(Omnibus):		0.000	Jar	que-Be	ra (JB):		22.098
Skew:		-0.019	Pro	b(JB):	9 9 - 3000 BS	1	.59e-05
Kurtosis:		3.137	Con	d. No.			724.

Jinja

		0LS Regr	ression	Result	s		
Dep. Variable:	pm2_5_	calibrated_va	alue F	 R−squar	 ed:		0.072
Model:			OLS A	Adj. R-	squared:		0.072
Method:		Least Squa	ares F	-stati	stic:		511.8
Date:		Sat, 15 Jul 2	2023 F	Prob (F	-statistic	:):	0.00
Time:		00:55	5:54 L	_og-Lik	elihood:		-16648.
No. Observations	:	26	304 A	AIC:			3.331e+04
Df Residuals:		26	5299 E	BIC:			3.335e+04
Df Model:			4				
Covariance Type:		nonrob	oust				
	coef	std err	=====	t	P> t	[0.025	0.975]
const	5.8014	0.074	77.89	 96	0.000	5.655	5.947
precipitation	-0.0307	0.003	-11.46	51	0.000	-0.036	-0.025
humidity	-1.4058	0.047	-30.01	11	0.000	-1.498	-1.314
temp_mean	-0.0771	0.002	-41.47	77	0.000	-0.081	-0.073
wind_speed	0.0421	0.003	16.72	22	0.000	0.037	0.047
 Omnibus:		520.733	Durb	in–Wats	====== on:		0.390
Prob(Omnibus):		0.000	Jarqu	ue-Bera	(JB):	1	129.053
Skew:		-0.014	Prob	(JB):		6.	75e-246
Kurtosis:		4.015	Cond.	No.			710.

GENERALISED LINEAR REGRESSION MODEL WITH LINKED GAMMA FUNCTION

Kampala

Generalized Linear Model Regression Results

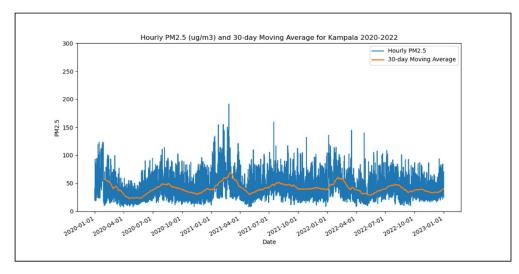
Dep. Variable:	pm2_5	_calibrated_	value No	. Observati	ons:	26304
Model:			GLM D	f Residuals:		26299
Model Family:			Gamma D	f Model:		4
Link Function:			log S	cale:		0.17188
Method:			IRLS L	g-Likelihoo	d:	-1.0879e+05
Date:		Sat, 15 Jul	2023 De	eviance:		4209.6
Time:		01:	01:45 P	earson chi2:		4.52e+03
No. Iterations	::		100 P	seudo R-squ.	(CS):	0.1837
Covariance Typ	e:	nonr	obust			
=========	coef	std err		z P> z	[0.025	0.975]
Intercept	-5.194e+10	6.09e+10	-0.85	0.394	-1.71e+11	6.74e+10
precipitation	-0.0147	0.002	-6.72	0.000	-0.019	-0.010
humidity	-1.8237	0.042	-43.16	0.000	-1.907	-1.741
temp_mean	-0.0915	0.002	-47.30	0.000	-0.095	-0.088
wind_speed	-0.3901	0.010	-39.23	0.000	-0.410	-0.371
intercept	5.194e+10	6.09e+10	0.85	0.394	-6.74e+10	1.71e+11

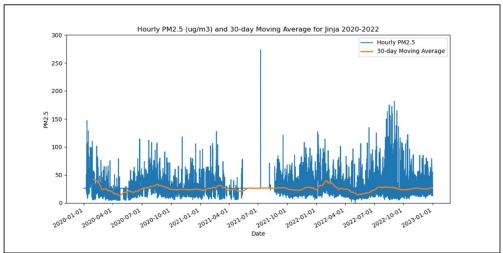
Jinja

Generalized Linear Model Regression Results

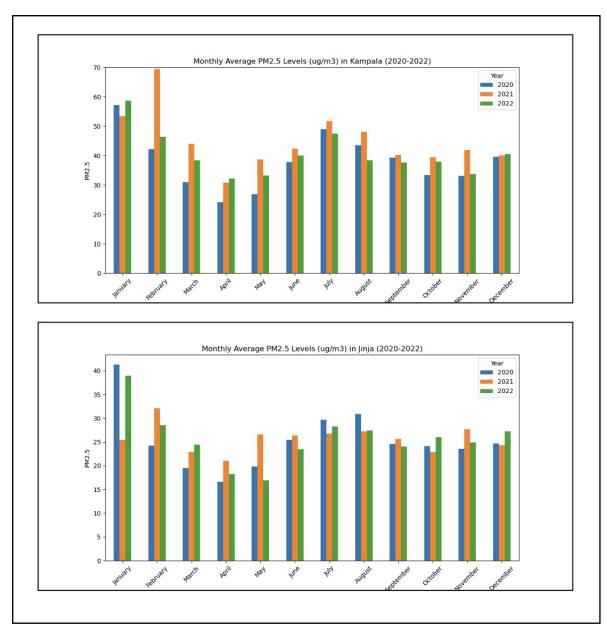
Dep. Variable:	pm2 5	calibrated v	alue No.	Observation	ns:	26304
Model:		_	GLM Df	Residuals:		26299
Model Family:		G	amma Df	Model:		4
Link Function:	:		log Sca	ile:		0.25885
Method:			IRLS Log	-Likelihood	l:	-99924
Date:		Sat, 15 Jul	2023 Dev	iance:		5537.1
Time:		01:0	1:02 Pea	rson chi2:		6.80e+03
No. Iterations	5:		100 Pse	eudo R-squ.	(CS):	0.06918
Covariance Typ	oe:	nonro	bust			
	coef	std err	z	P> z	[0.025	0.975]
Intercept	-1.282e+10	7.47e+10	-0.172	0.864	-1.59e+11	1.34e+11
precipitation	-0.0316	0.003	-10.570	0.000	-0.037	-0.026
humidity	-1.3988	0.052	-26.744	0.000	-1.501	-1.296
temp_mean	-0.0821	0.002	-39.550	0.000	-0.086	-0.078
wind_speed	0.0346	0.003	12.314	0.000	0.029	0.040
intercept	1.282e+10	7.47e+10	0.172	0.864	-1.34e+11	1.59e+11

Results

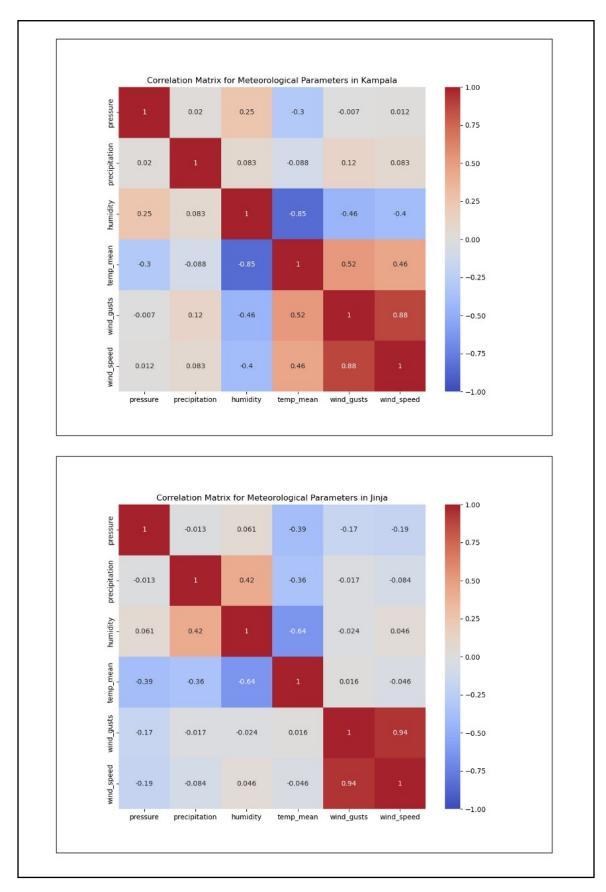




 \mathbf{SF} 4: Hourly $PM_{2.5}$ (µg/m³) and 30-day Moving Average for Kampala and Jinja from 2020-2022



SF 5:Temporal Variation of $PM_{2.5}$ (µg/m³) in Kampala and Jinja from 2020 to 2022



SF6: Correlation Matrix for Meteorological Parameters in Kampala and Jinja