

Potential for well-calibrated air sensors to enhance utility of satellite retrievals for air quality management

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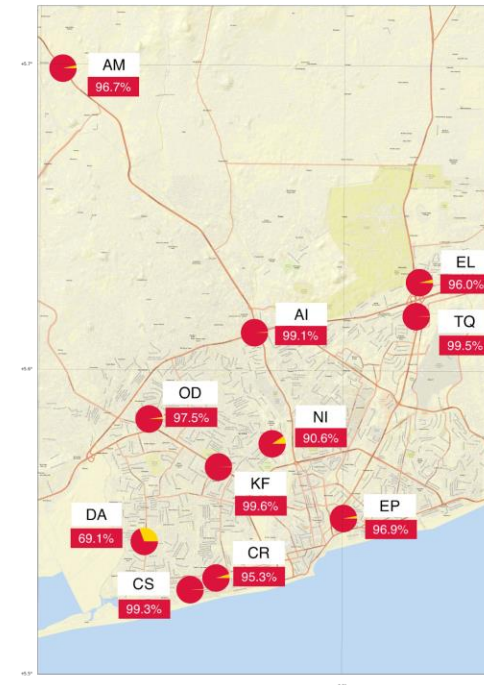
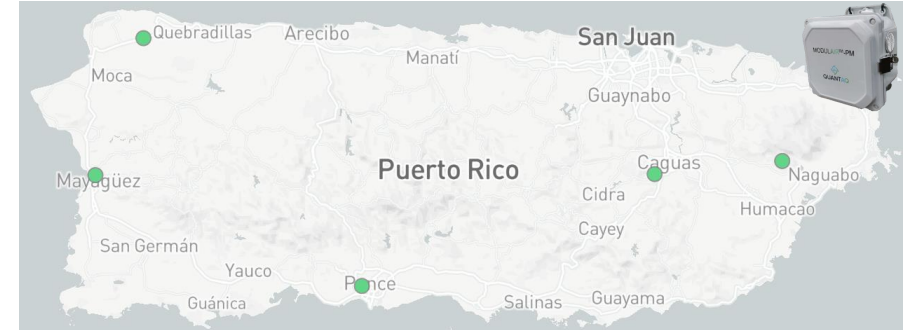
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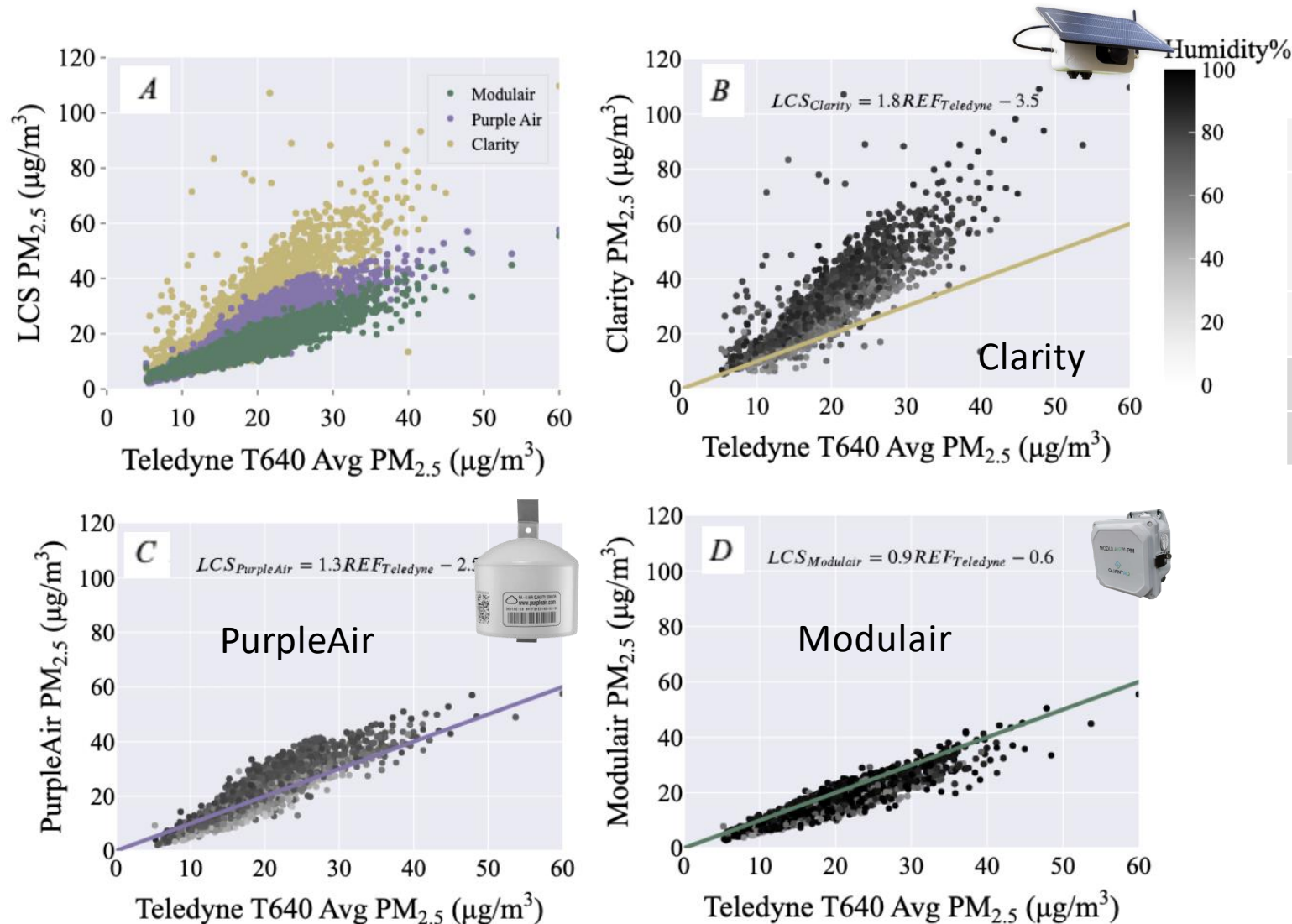
 COLUMBIA UNIVERSITY
DATA SCIENCE INSTITUTE



Extracting useful, actionable data from air sensors to close air pollution data gaps around the world



Research projects – Sensor performance between different vendors against reference PM2.5



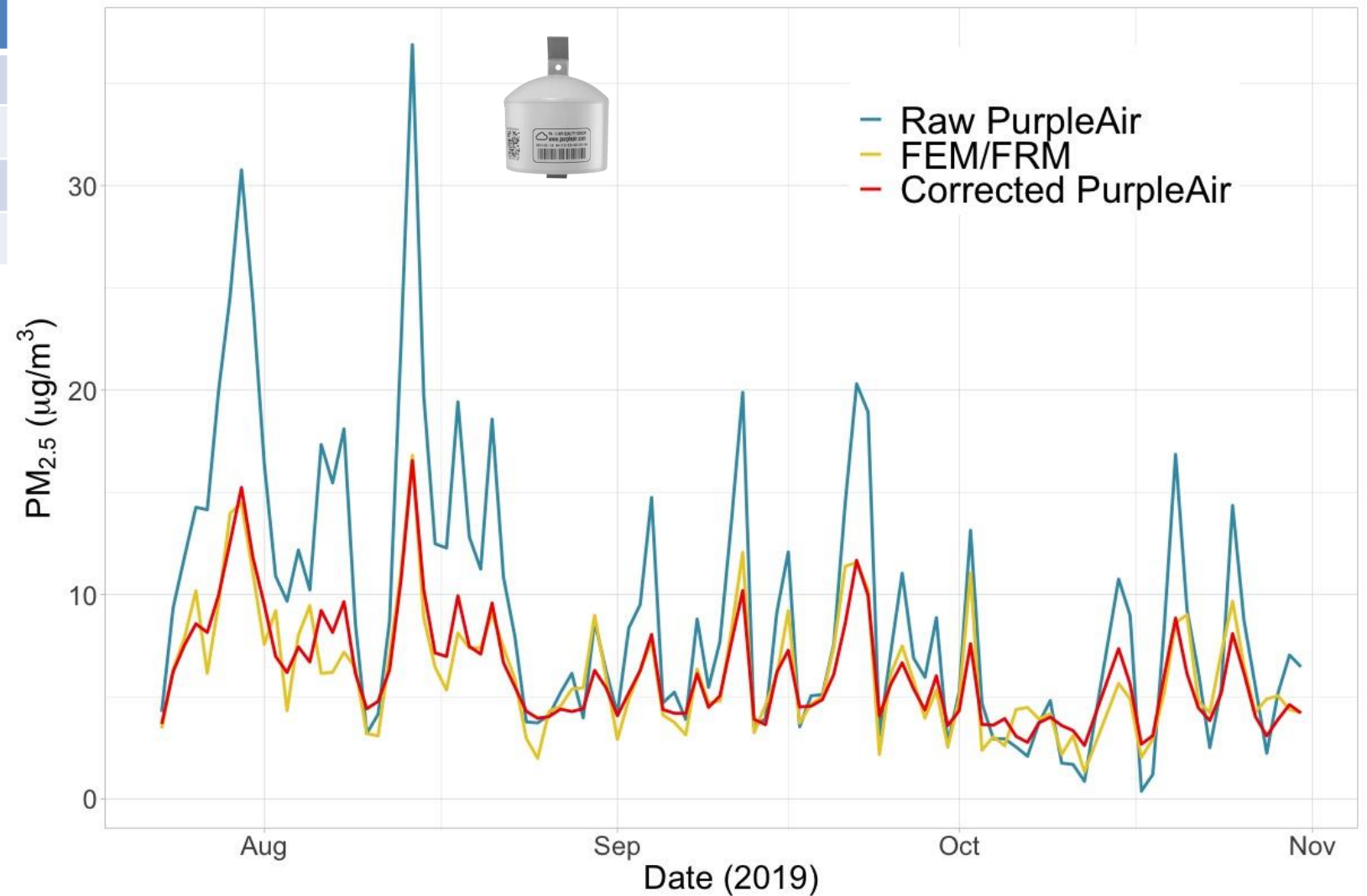
Model	Purple Air		Clarity		Modulair	
	R^2	MAE (μg/m ³)	R^2	MAE (μg/m ³)	R^2	MAE (μg/m ³)
Manufacturer-Reported	0.82	4.54	0.69	13.68	0.84	3.04
MLR	0.87	1.96	0.74	2.49	0.86	2.15
GMR	0.89	1.76	0.79	2.27	0.87	1.89

Set aside some fraction of the data to as “training data” to develop “correction factor”



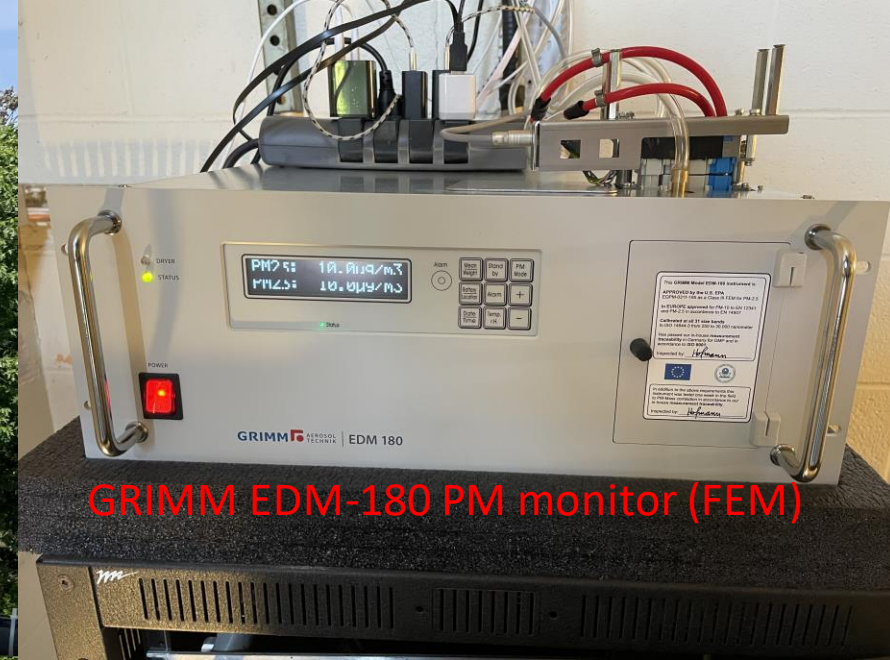
NYC calibrations at Near-Roadway site (Queens College)

	Spearman r	MAE ($\mu\text{g m}^{-3}$)
Raw-NR	0.801	4.02
Corr-NR	0.817	1.83
Raw-QB	0.757 (0.88)	5.01
Corr-QB	0.849 (0.87)	1.65

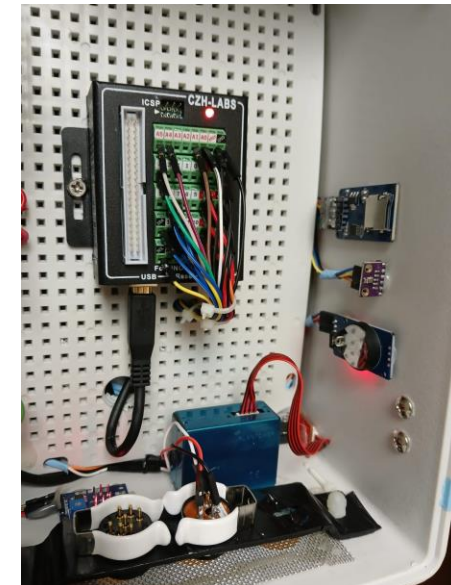


Columbia-LDEO Air Sensors lab

Sensor vs FEM intercomparison site



GRIMM EDM-180 PM monitor (FEM)



“homemade”
air sensors:
PM, tVOCs, NO₂

Satellite data: what role for air sensors?

- **The problem**: satellite estimates provide excellent spatial coverage of pollutants at decent resolution (1 km² daily), but measure the entire atmospheric column, rather than “breathing level”
- **One solution**: geophysical model to convert AOD to surface PM_{2.5} (van Donkelaar, Martin, etc)

$$PM_{2.5} = \eta(x, y, t) \times AOD_{BE}$$

GEOS-Chem

- **Another solution**: use statistical models to develop an empirical relationship between surface PM_{2.5} and column AOD



$$PM_{2.5} = f(AOD, T, Pr, NO_2, CO)$$



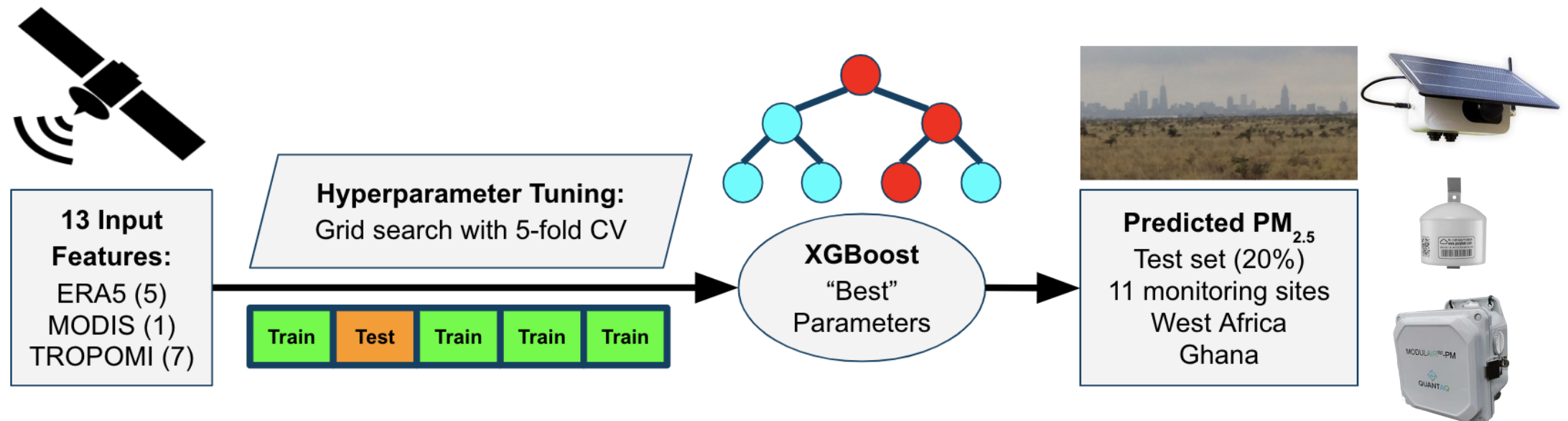
High Spatiotemporal Resolution Modeling of PM_{2.5} in West Africa Using Satellite Data and Machine Learning

Goal: Estimate PM_{2.5} at high spatiotemporal resolution (1 km², daily) in West Africa over the past 2 decades

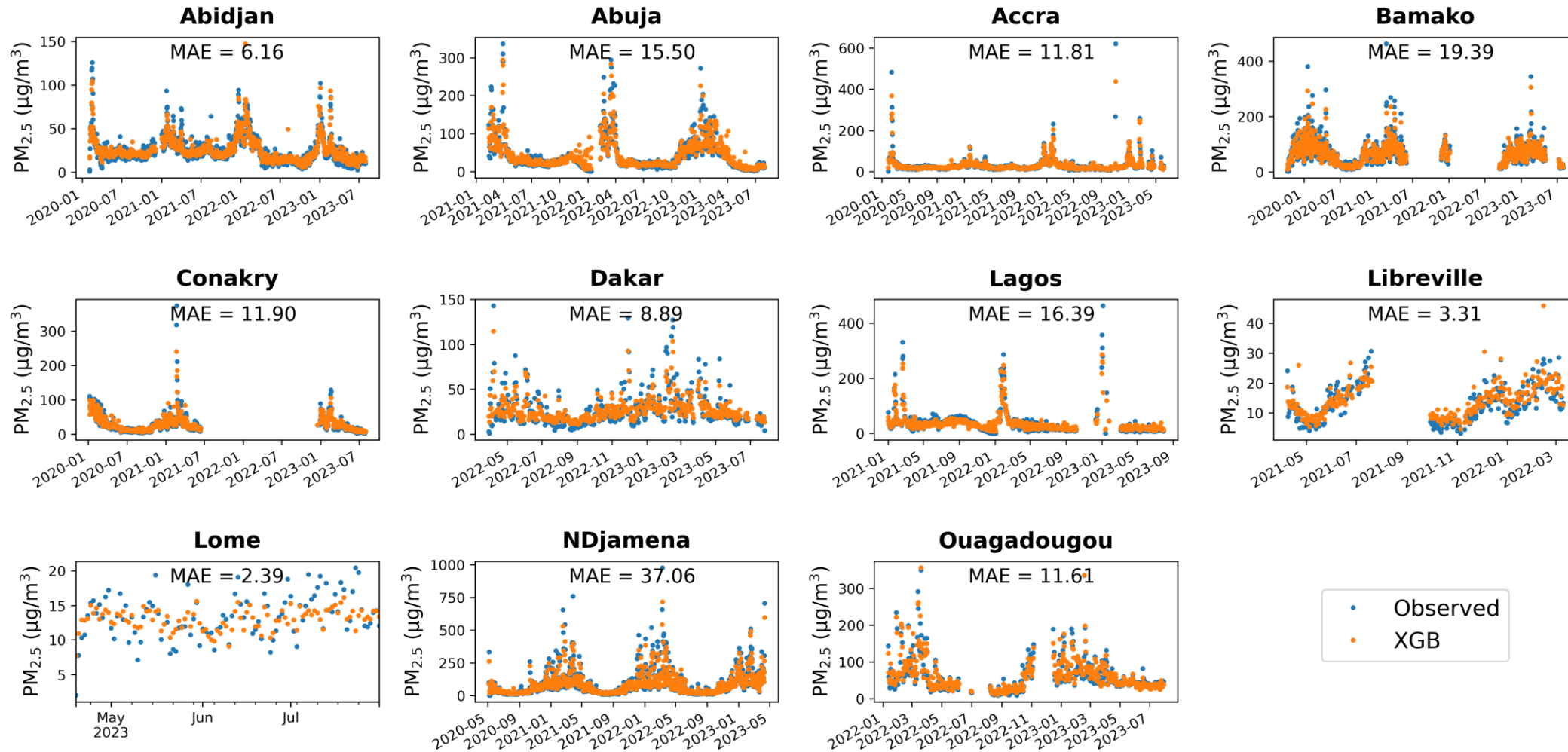
Data & Methods: Trained, tested, and fine-tuned a machine learning (XGBoost) model:

- ✓ PM_{2.5} from 11 U.S. Embassies & Consulates monitoring sites
- ✓ AOD from MODIS MAIAC satellite retrievals
- ✓ 5 meteorological features from ERA5
- ✓ 7 trace gas or aerosol features from TROPOMI satellite retrievals

Applications: Epidemiological, policy, and environmental justice studies (Ghana)

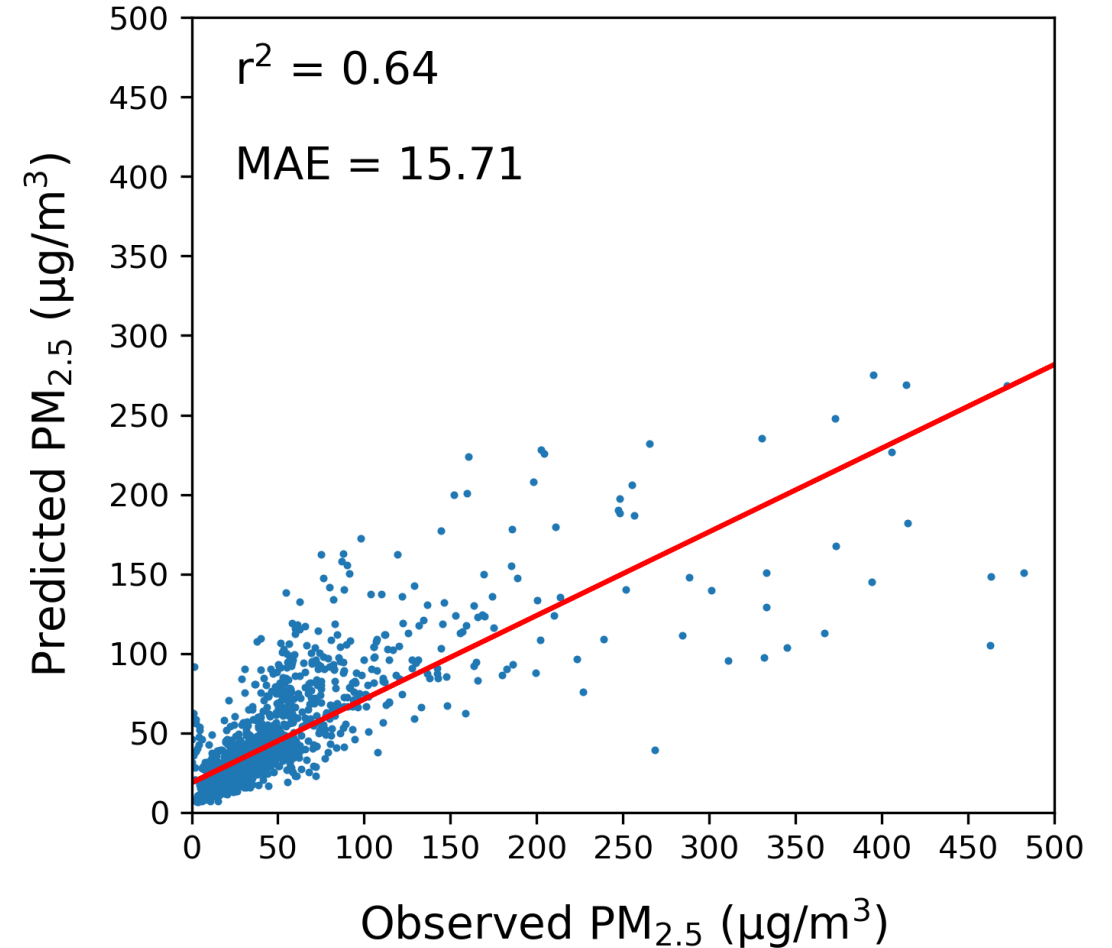
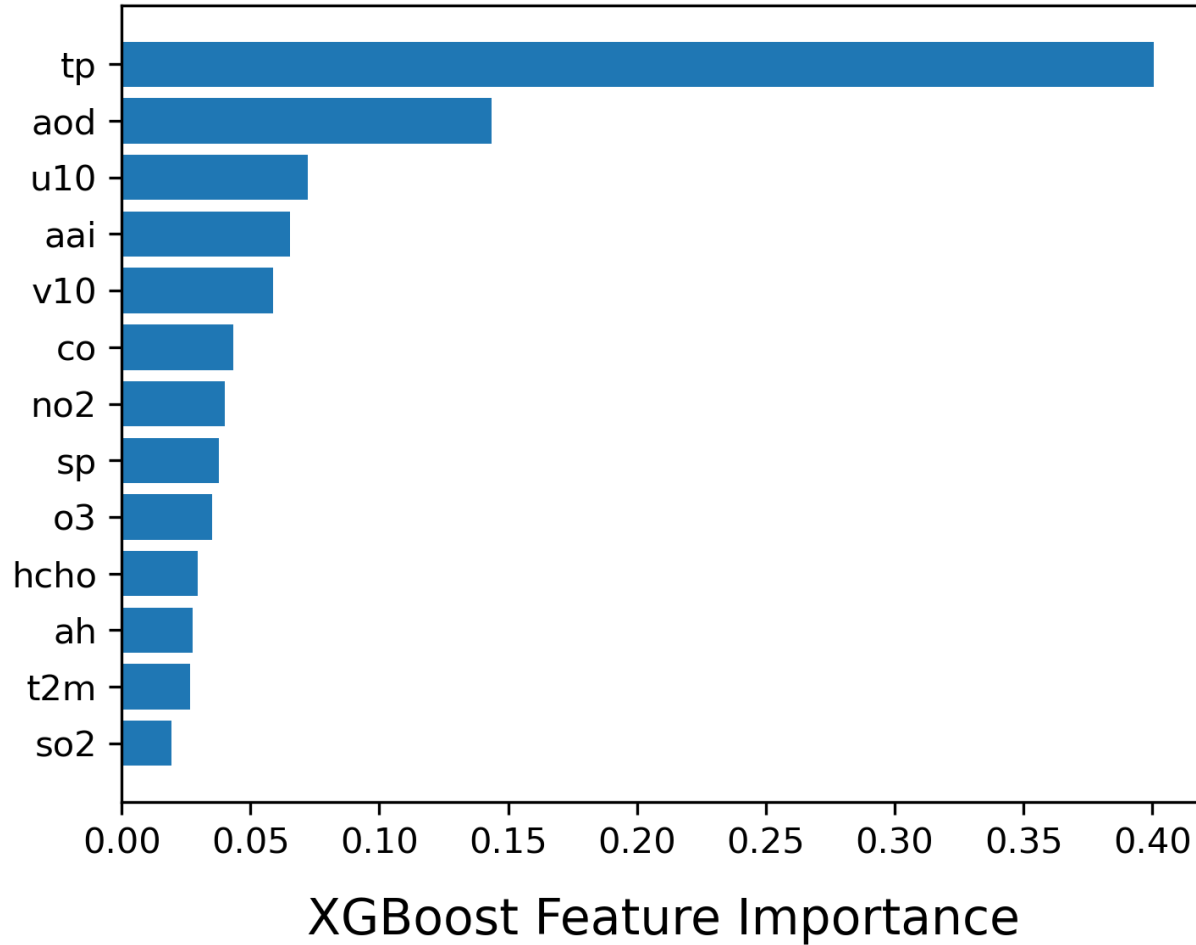


Predicted ML-satellite PM2.5 versus observations

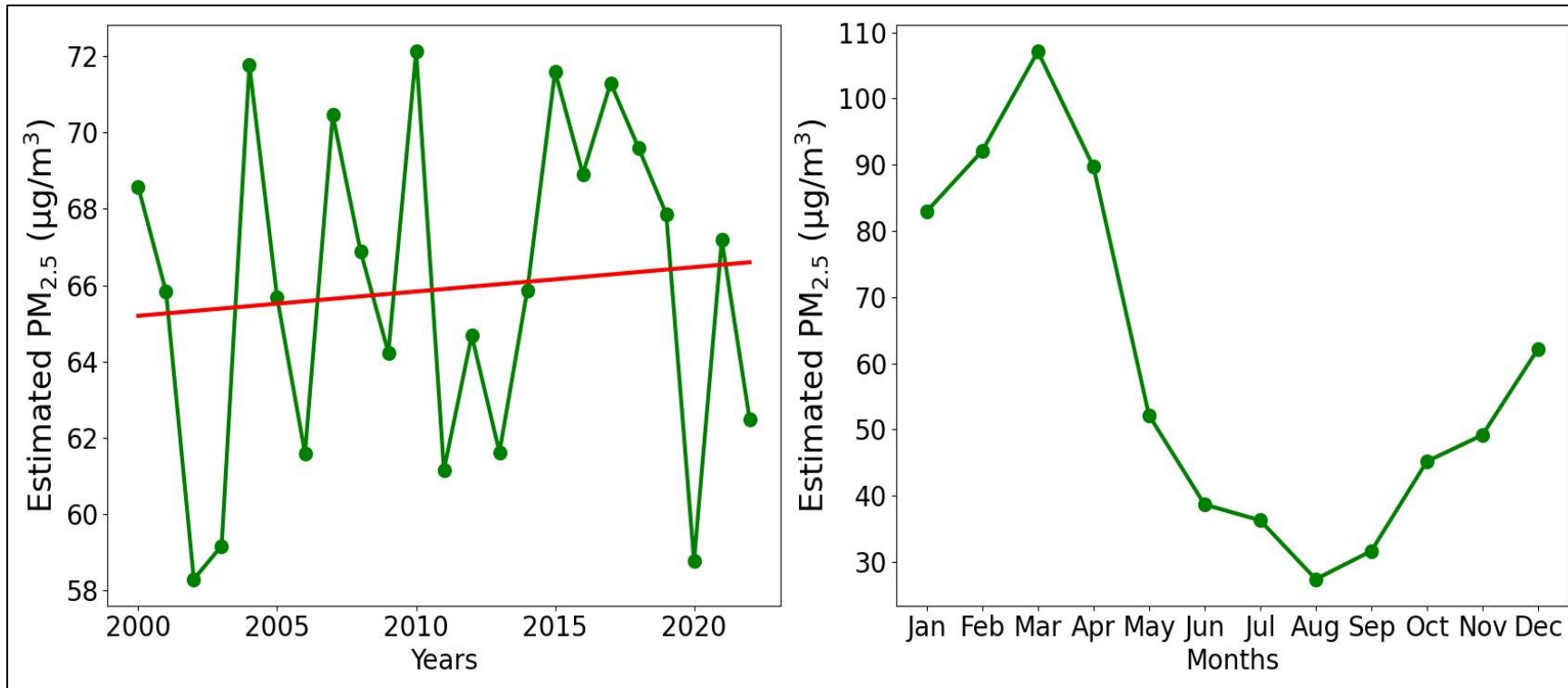


- $R^2 \sim 0.7$
- Most important features: AOD, precipitation, UVAI, windspeed

Predicted ML-satellite PM_{2.5} versus observations (daily)

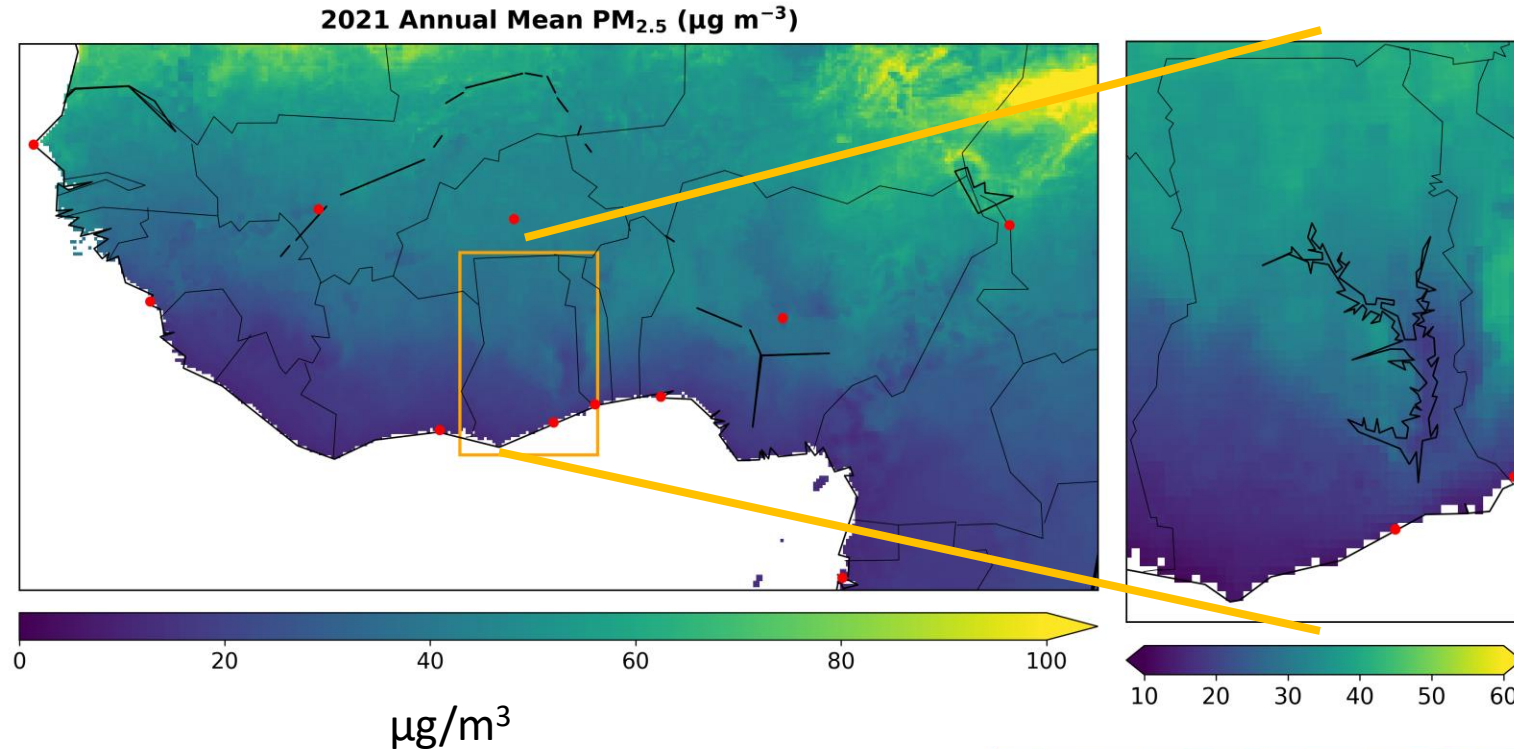


Results from Burkina Faso



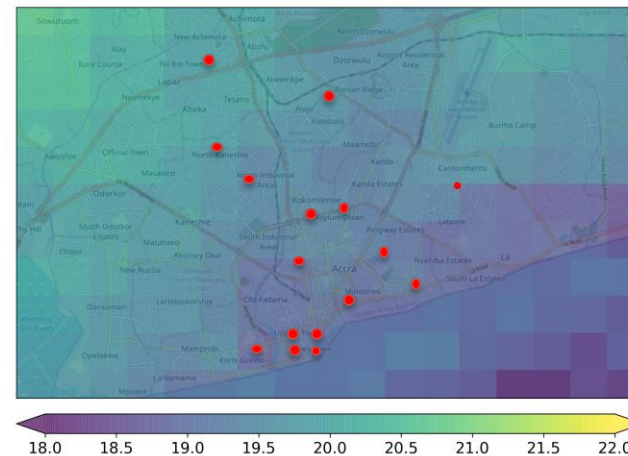
- XGBoost model has $r^2 = 0.87$ and nRMSE $\sim 20\%$
- Evidence of slight increases in PM_{2.5} since 2000 (beginning of satellite record), with substantial interannual variability

1 km² spatial resolution



- Can apply statistical relationship back through the satellite record to get a long term (2000-present), daily, 1km² dataset

Applications:
Epidemiological, policy,
and environmental
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Summary

- Growing hybrid observational networks of surface PM_{2.5} over Africa and other traditionally data-sparse areas can be used to convert columnar AOD to “breathing level” PM_{2.5}
 - Recommend that this is applied and benchmarked more in the Global South as novel datasets are coming online
- We demonstrate satellite-ML predictions of PM_{2.5}, trained against air sensor data, with high r^2 (>0.7) and low error (20%)
 - Suitable for use in long-term trends, policy evaluation, epidemiology