

City Scanner Workshop



senseable
city lab.

Workshop Agenda + Goals

Agenda

- CityScanner Recap
- Historical + environmental use case context on The Bronx
- Analysis + Coding activities

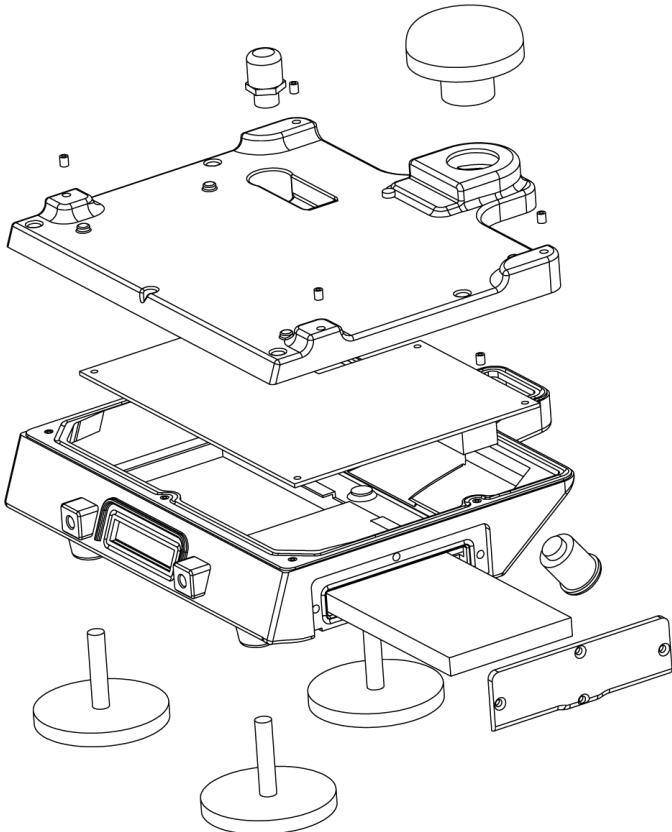
Goals

- Understanding hyperlocal air quality sensing
- Perform time series + hotspot analysis
- Make maps with data



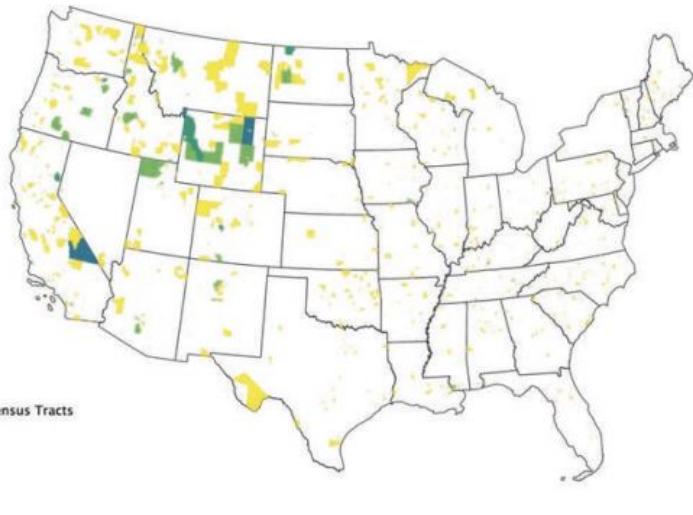
Can we turn urban vehicles into
sensing platforms?

City Scanner

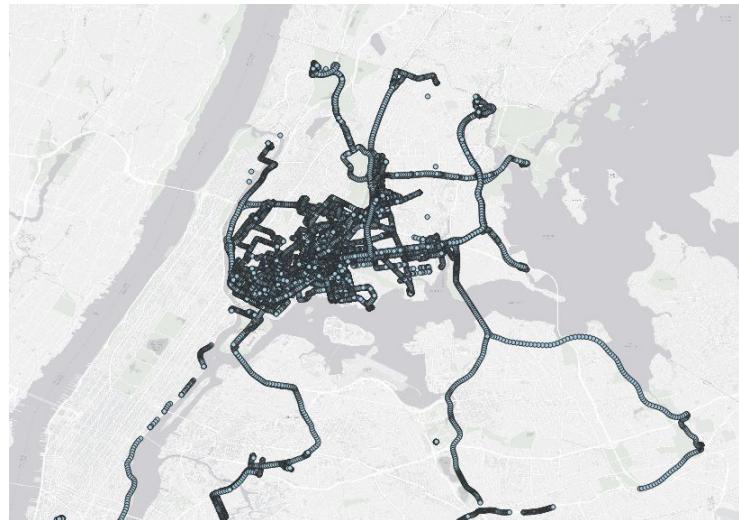


- Location
- Air quality (PM, CO, NOx)
- Temperature & Humidity
- Noise

Stationary vs. Mobile Sensors



EPA Monitors that report PM_{2.5} from 2015 to Feb 22 2020 per census tract in the US ([deSouza and Kinney 2021](#))



Space coverage achieved with five city scanner sensors deployed in the Bronx for 3 months

A world map with a halftone dot pattern. Six specific locations are highlighted with green text labels and small dots:

- Sparwood 2019
- New York City 2020
- Cambridge 2017
- Stockholm 2020
- Beirut 2021
- Oskemen 2021

The labels are positioned near their respective project locations: Sparwood in British Columbia, New York City in the United States, Cambridge in England, Stockholm in Sweden, Beirut in Lebanon, and Oskemen in Kazakhstan.

Sparwood
2019

New York City
2020

Cambridge
2017

Stockholm
2020

Beirut
2021

Oskemen
2021

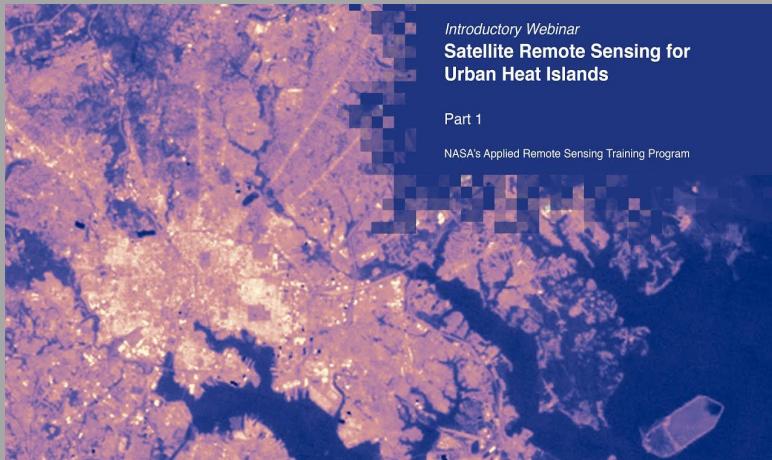
Environmental Sensing

Context + Background

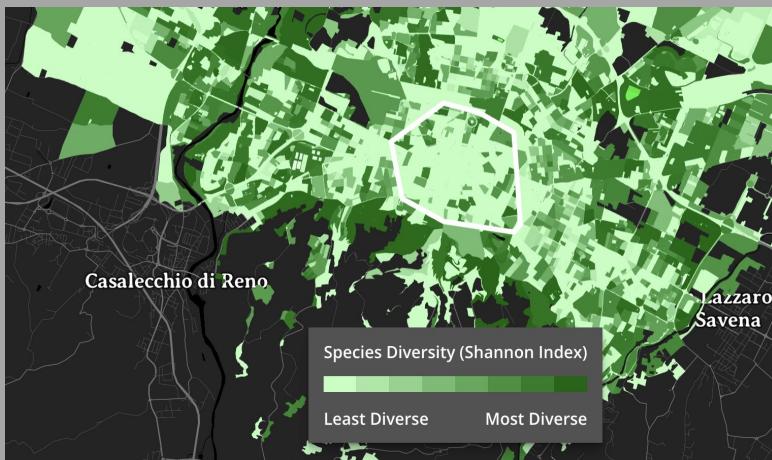
ENVIRONMENTAL SENSING 101

What information can we gather about our environment with different sensors?

Heat, noise, air quality, temperature humidity, soil health, water pollutants, tree health, biodiversity, and more!

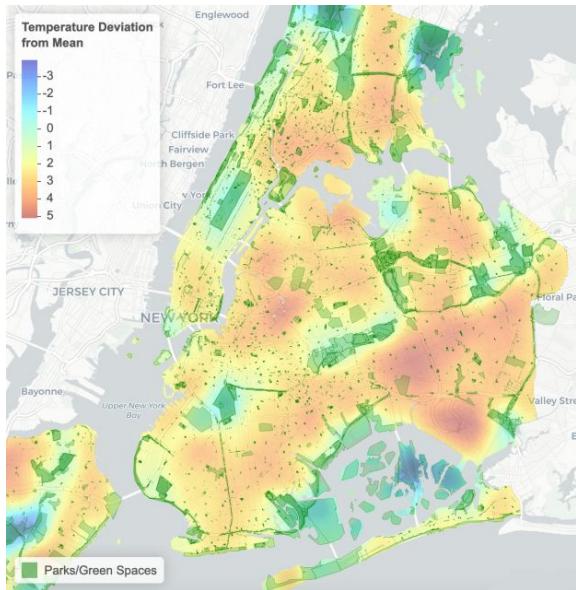


NASA ARSET

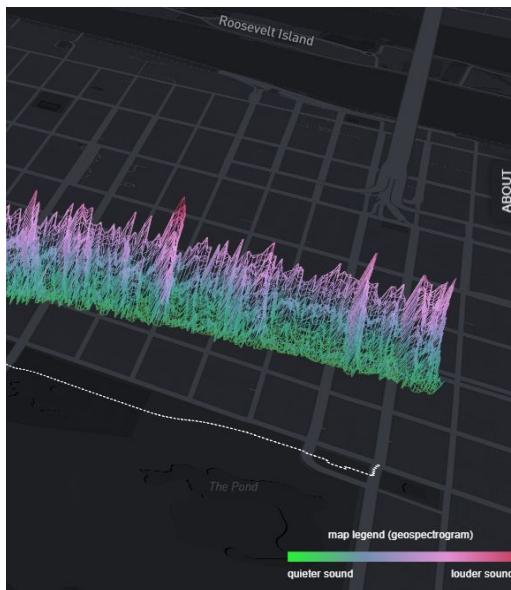


DiversiTTree

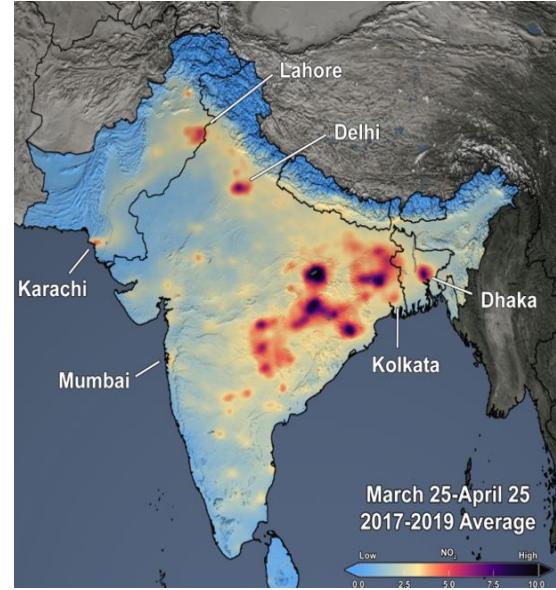
Environmental Sensing: Heat, Noise, Air Quality



<https://news.climate.columbia.edu/2021/08/26/study-maps-urban-heat-islands-with-focus-on-environmental-justice/>



<https://senseable.mit.edu/sonic-cities/>



<https://aura.gsfc.nasa.gov/airquality.html>

Air Quality Sensing: Why?



The Great Smog 1952
(<https://www.britannica.com/event/Great-Smog-of-London>)



Los Angeles Smog
(<https://www.britannica.com/science/smog#ref16459>)

Air Quality Sensing: Why?

4.2 million per year vs. 6.2 million

91% of world population lives in places exceeding WHO AQ Standards

RESEARCH ARTICLE | SOCIAL SCIENCES | FULL ACCESS



Half of US population exposed to adverse lead levels in early childhood

Michael J. McFarland   , Matt E. Hauer  , and Aaron Reuben [Authors Info & Affiliations](#)

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THIS ARTICLE HAS BEEN UPDATED

Air Quality Index: Pollutants

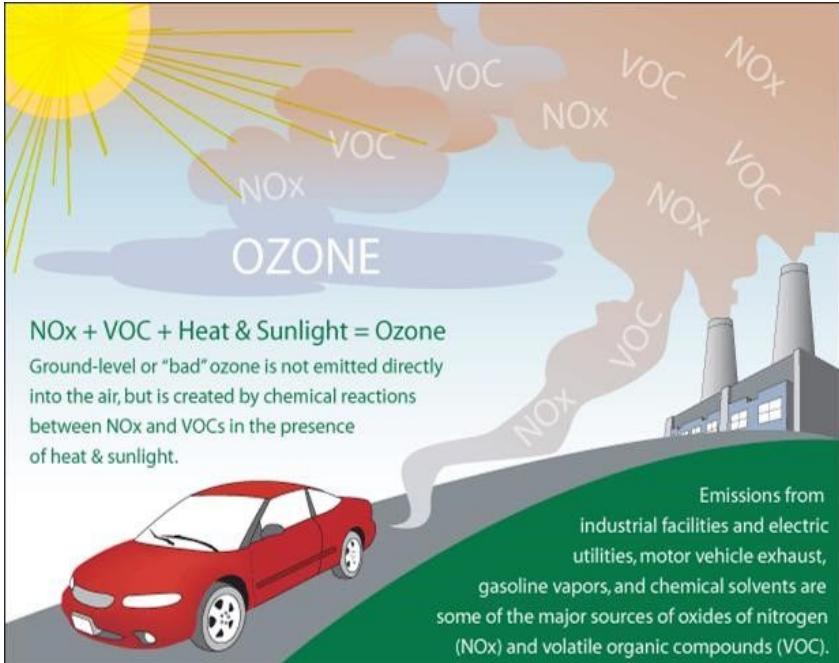


Six criteria pollutants regulated by EPA

- Ground-level ozone
- Particulate matter
- Carbon monoxide
- Lead
- Sulfur dioxide
- Nitrogen dioxide

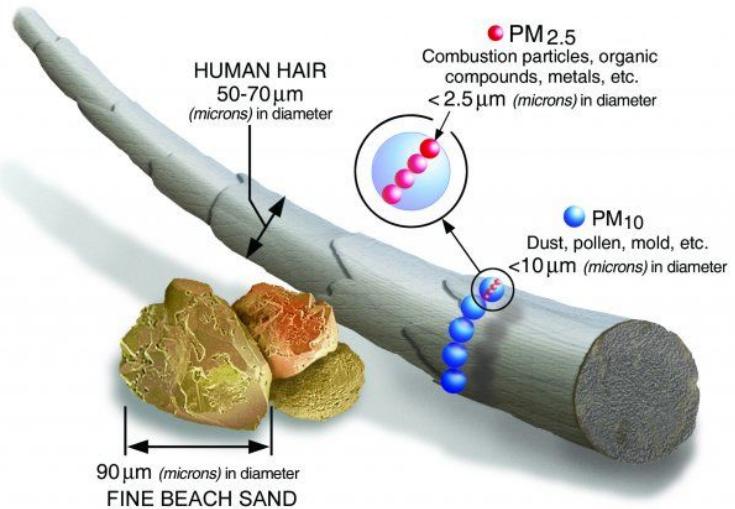
[United States Environmental Protection Agency](http://www.epa.gov)

Air Quality Index: Pollutants



<https://www.epa.gov/ground-level-ozone-pollution/ground-level-ozone-basics#formation>

Air Quality Index: PM and NO₂

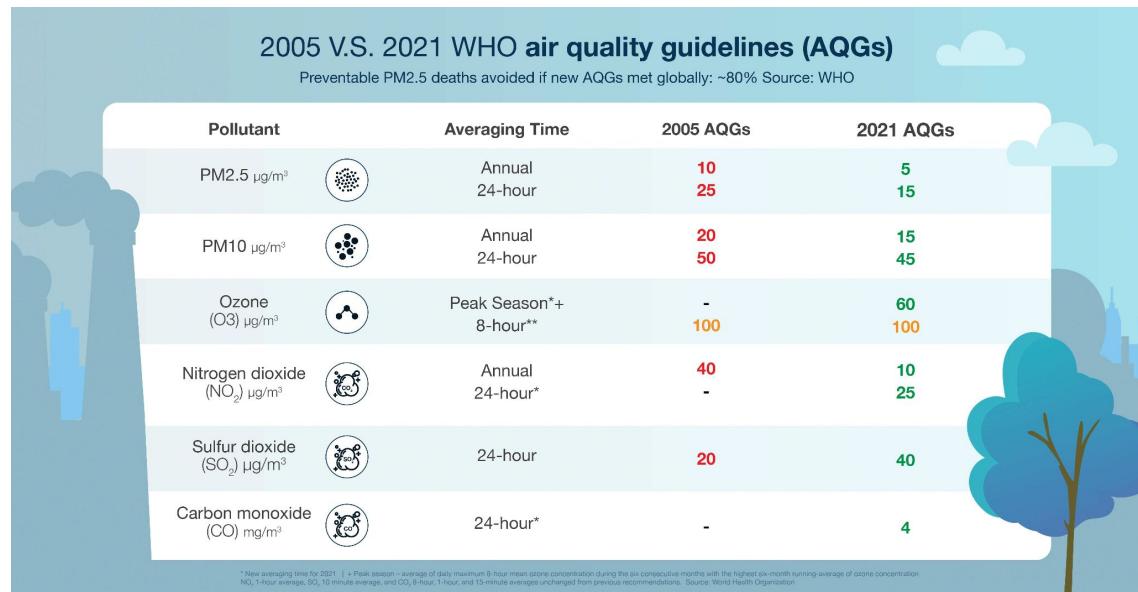


<https://www.epa.gov/pm-pollution/particulate-matter-pm-basics#PM>



<https://phys.org/news/2018-03-german-deaths-nitrogen-dioxide.html>

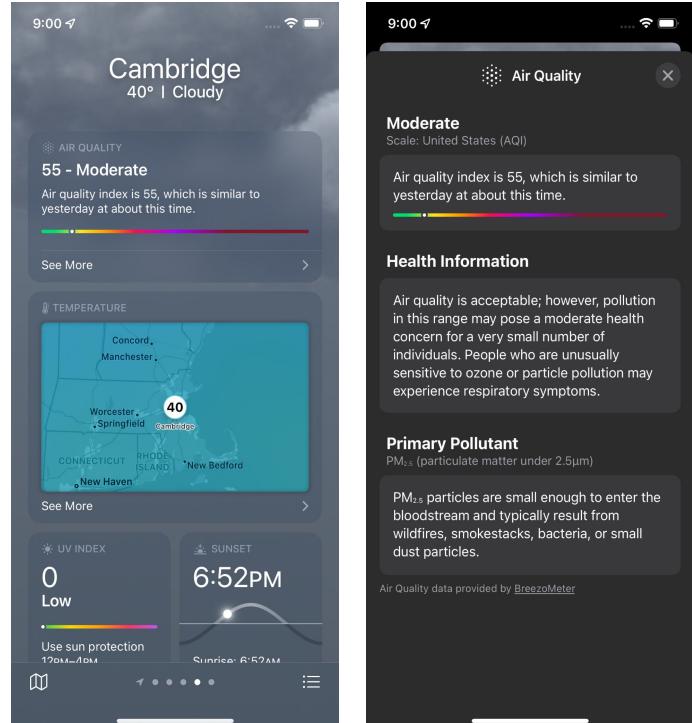
World Health Organization Guidelines



Air Quality Index

Air Quality Index		
AQI Category and Color	Index Value	Description of Air Quality
Good Green	0 to 50	Air quality is satisfactory, and air pollution poses little or no risk.
Moderate Yellow	51 to 100	Air quality is acceptable. However, there may be a risk for some people, particularly those who are unusually sensitive to air pollution.
Unhealthy for Sensitive Groups Orange	101 to 150	Members of sensitive groups may experience health effects. The general public is less likely to be affected.
Unhealthy Red	151 to 200	Some members of the general public may experience health effects; members of sensitive groups may experience more serious health effects.
Very Unhealthy Purple	201 to 300	Health alert: The risk of health effects is increased for everyone.
Hazardous Maroon	301 and higher	Health warning of emergency conditions: everyone is more likely to be affected.

Air quality index ([United States Environmental Protection Agency](#))



Air quality index shown in iPhone weather app

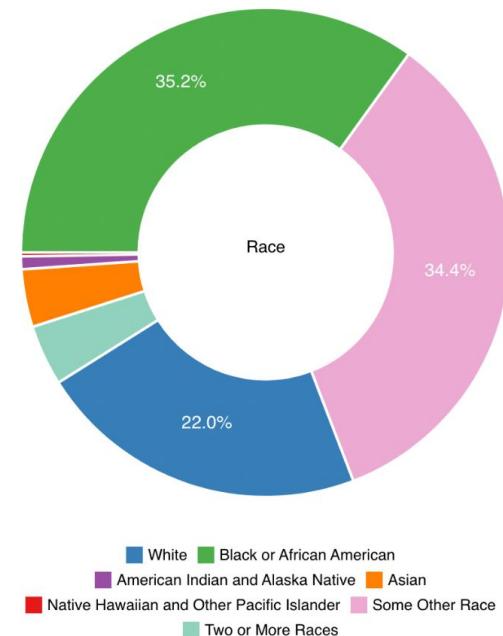
The Bronx

Historical + Environmental Context

The Bronx: Overview + Demographics



US Census 2019 ACS 5-Year Survey (Table B03002)

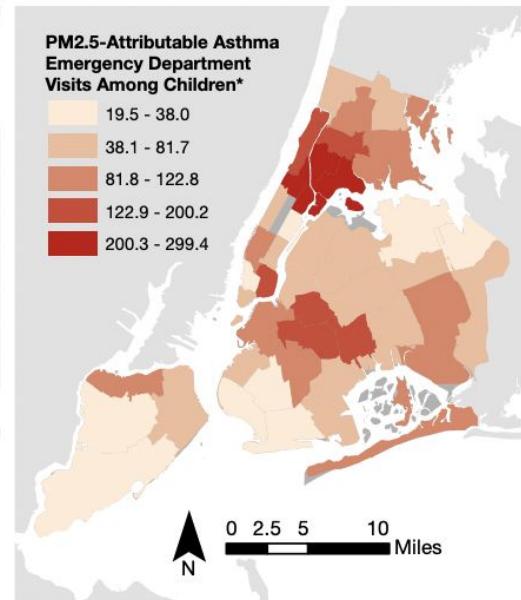
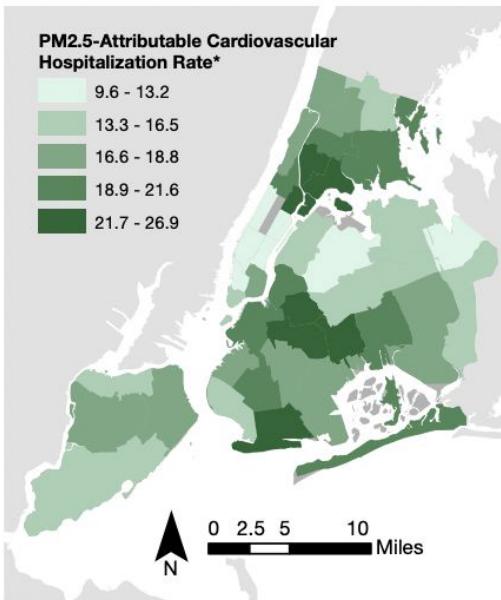
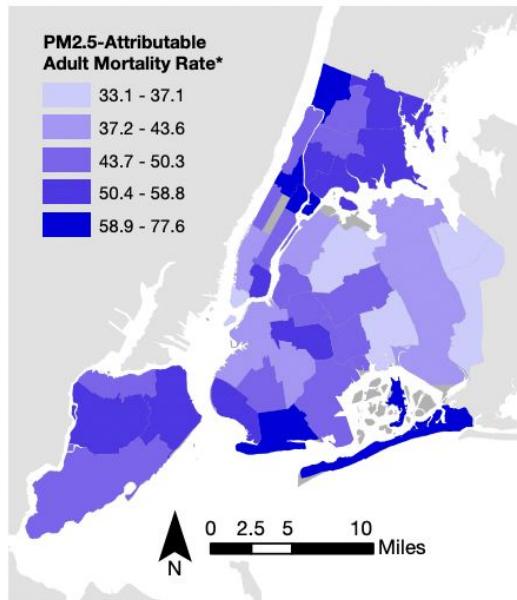


The Bronx: Context



The Bronx: Environmental Issues

Mortality and morbidity from selected conditions due to PM2.5 in New York City



* 2009-2011 Annual Average, Rate per 100,000 persons

The Bronx: Environmental Issues

Resources

- [NYCCAS Data](#)
- [South Bronx Environmental Health and Policy Study](#)
- [NYC Environmental and Health Portal](#)
- [New York Disadvantaged Communities Criteria](#)
- [Climate and Economic Justice Screening Tool](#)
- [NYC Environmental Justice Alliance](#)
- [NYC Community Health Profiles](#)
- [Potential Environmental Justice Areas in The Bronx](#)

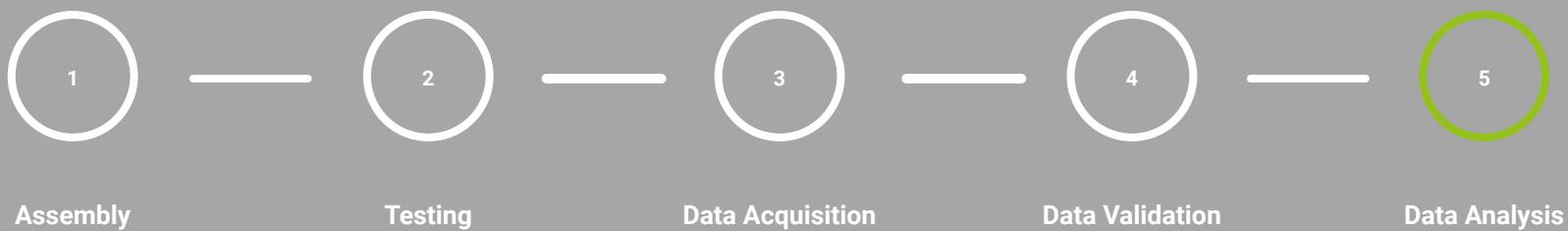
City Scanner

Data Collection Pipeline Tour

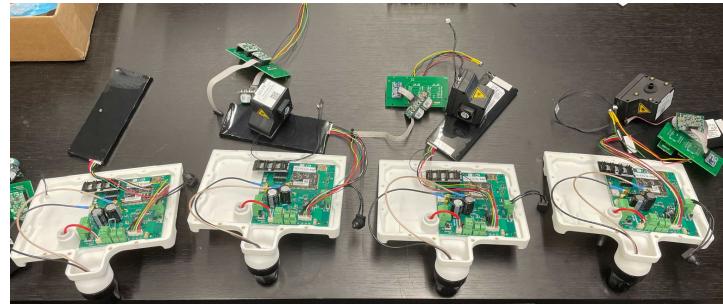
CITYSCANNER DATA COLLECTION PIPELINE TOUR

What does an environmental data collection pipeline look like in practice?

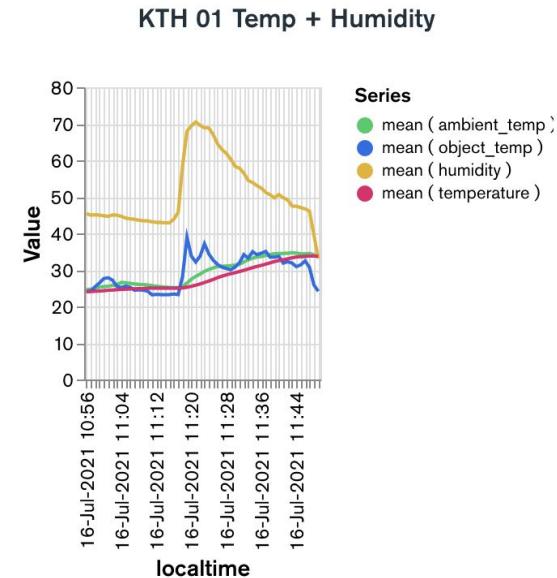
City Scanner: Data pipeline from hardware assembly to dynamic maps



Part 1: Assembly



Part 2: Testing



Part 3: Data Acquisition



Part 4: Data Validation

```
print("Total invalid timestamps for the collecti
# of invalid timestamps for KTH01: 0
# of invalid timestamps for KTH02: 0
# of invalid timestamps for KTH03: 1
# of invalid timestamps for KTH04: 0
# of invalid timestamps for KTH05: 74
Total invalid timestamps for the collection: 75

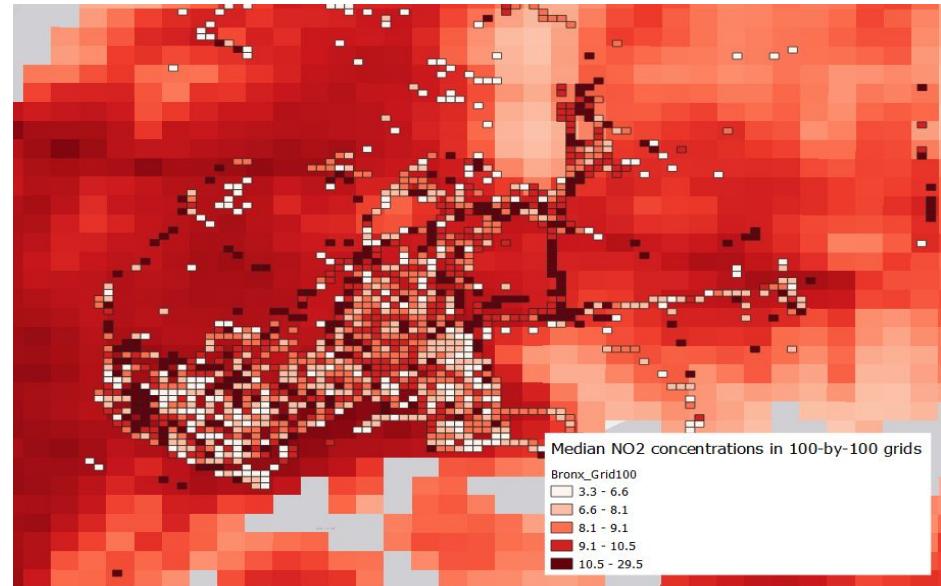
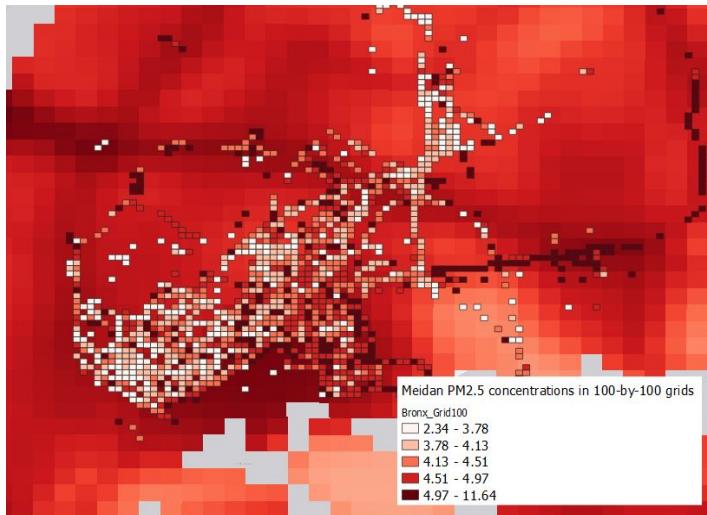
#find NAN/zero lat/lon values before filtering
##NOTE - this block must be run before data fil
zerolon1=(device1['longitude']==0).sum()
zerolat1=(device1['latitude']==0).sum()
print("Number of lat = 0 for KTH1:", zerolat1),
```

	ambientIR	gas_op1_r	gas_op1_w	gas_op2_r	gas_op2_w
--	-----------	-----------	-----------	-----------	-----------

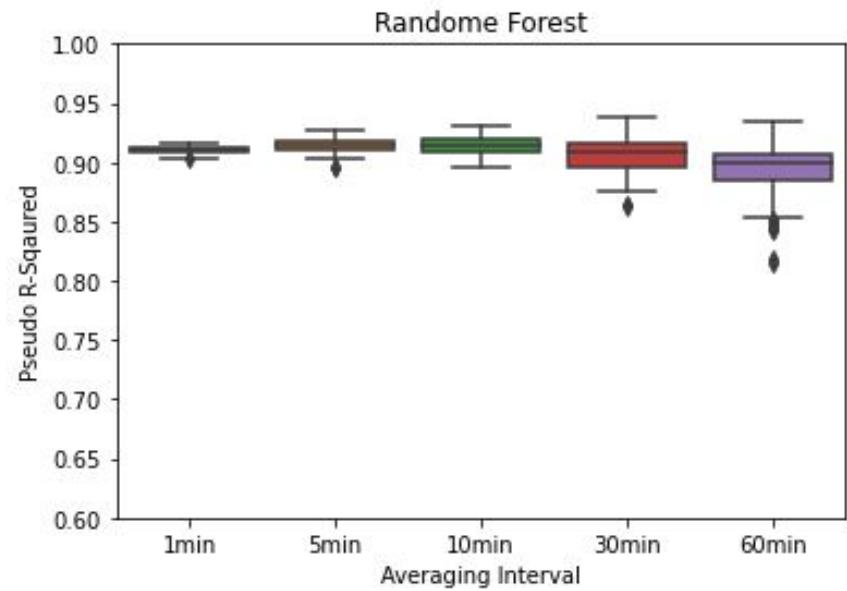
localtime					
-----------	--	--	--	--	--

2021-07-26 00:00:00+00:00	27.894032	278.498540	285.698994	252.880636	420.083036
2021-07-27 00:00:00+00:00	29.039576	279.636421	286.367491	243.619637	396.145507
2021-07-28 00:00:00+00:00	24.004650	384.980391	388.208495	350.313130	528.279259
2021-07-29 00:00:00+00:00	23.968180	341.854801	346.847965	322.335673	425.801422
2021-07-30 00:00:00+00:00	21.251523	346.355689	352.861879	326.958869	473.290794
2021-07-31 00:00:00+00:00	21.646791	352.544802	352.500273	308.294471	776.294315

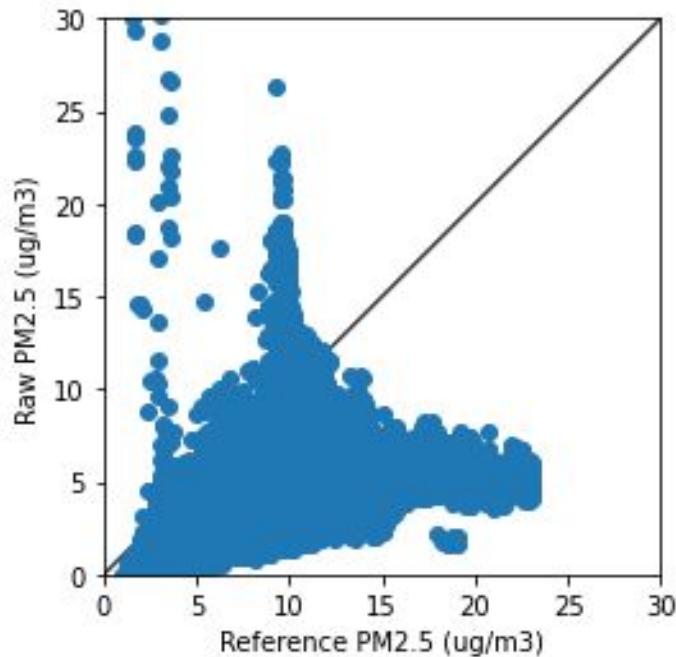
Part 5: Data Analysis



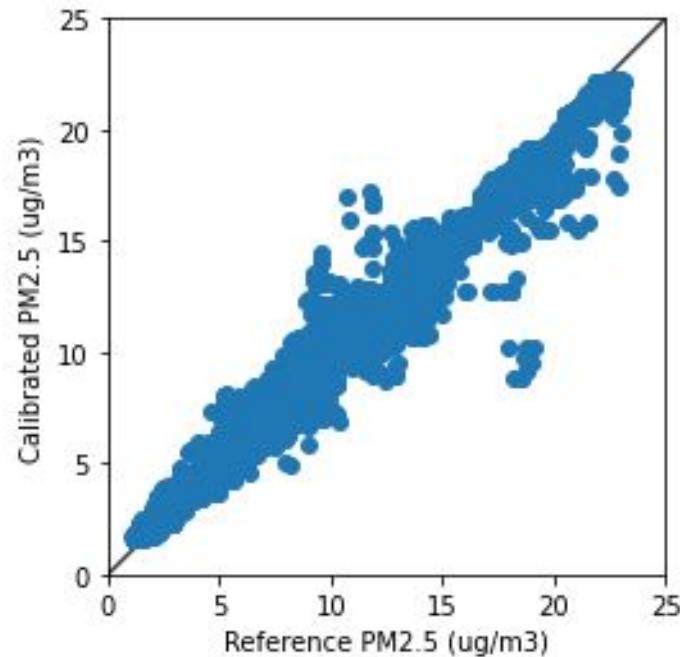
Analysis Method: Colocation + Calibration



Analysis Method: Colocation + Calibration



Before Calibration



After Calibration

Validation Method: Background Correction

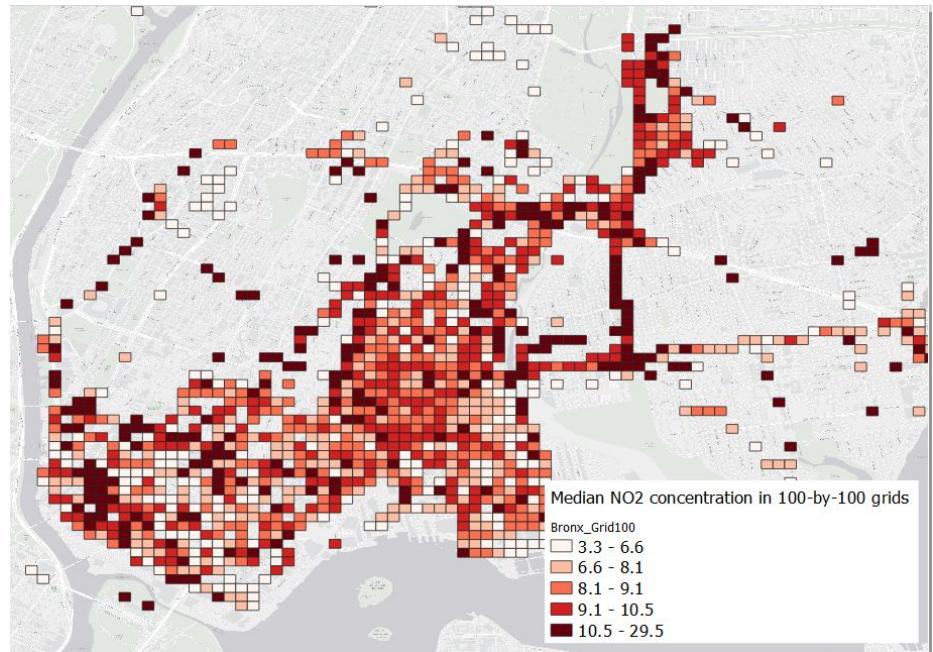
Background correction

- (Hourly) Multiplicative factor
- (Hourly) Lowest 10th percentile
- (Time series) Spline of minimums
- Background time-of-day correction

- Additive background correction factor

$$\text{PM}_{2.5,\text{norm } i} = \text{PM}_{2.5,\text{OPC } i} - \text{PM}_{2.5, \text{bkg},i} + \text{PM}_{2.5, \text{bkg,median}}$$

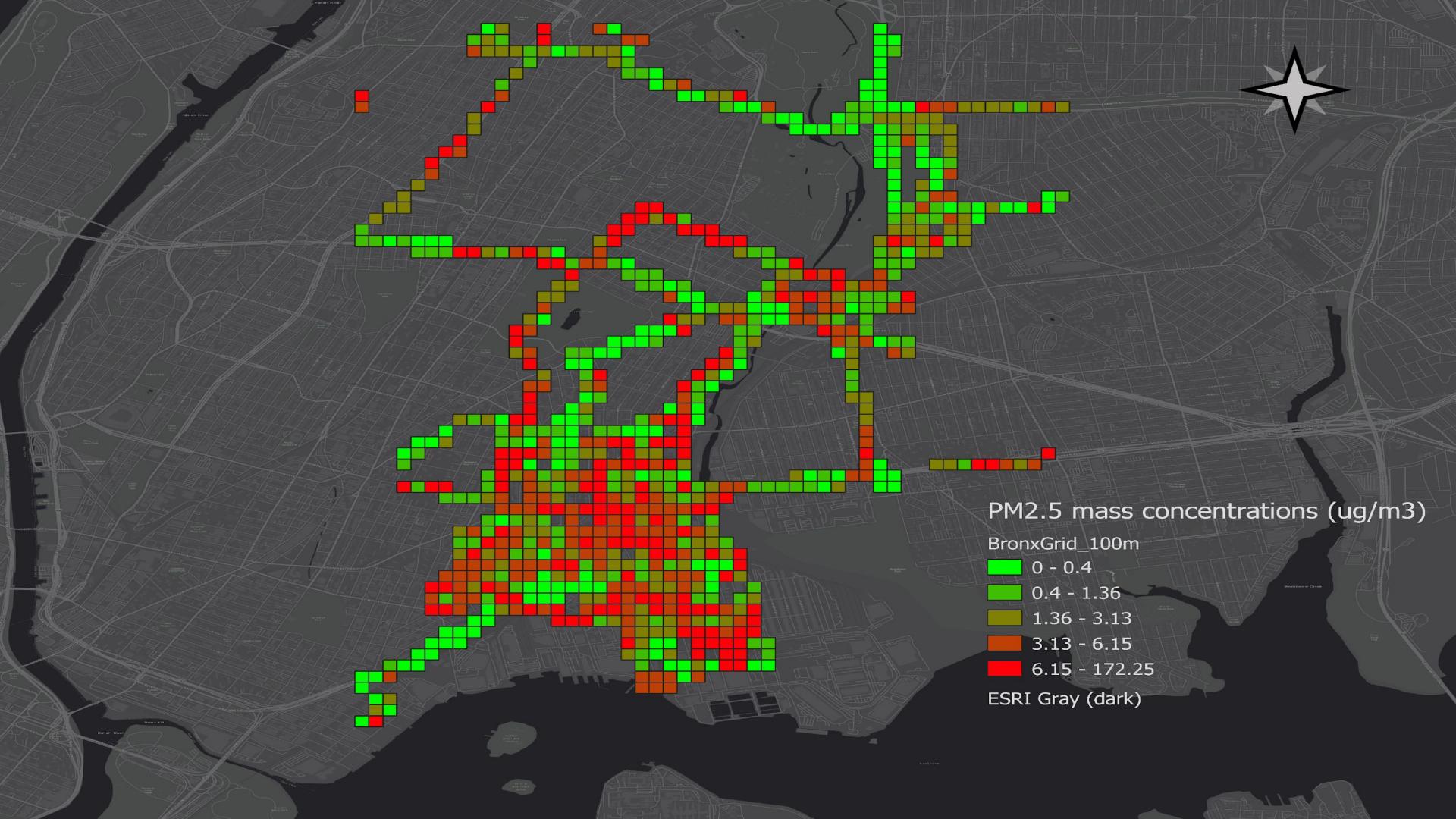
$$\text{PM}_{2.5,\text{norm } i} = \text{PM}_{2.5,\text{OPC, } i} \times \text{PM}_{2.5, \text{bkg,median}} / \text{PM}_{2.5, \text{bkg,i}}$$



CityScanner 2021 Data

Bronx 2021 Deployment

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z	AA	AB	AC	AD	AE	AF	AG	AH	AL	AI	AK	BL	AM	AN	AO	AP	AC	BR	AS	AT	BU
350030000	158-E9	40.80738	-73.94307	0.26	0.27	0.27	6	1	bin2	bin3	0	bin4	bin5	0	bin6	bin7	0	bin8																												
350030000	158-E9	40.80739	-73.94308	0.35	0.36	0.36	9	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.8074	-73.94307	0.32	0.33	0.33	8	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.8074	-73.94307	0.5	0.93	0.95	8	0	0	1	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80741	-73.94306	0.42	0.5	0.5	63	17	2	1	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80741	-73.94306	0.51	0.61	0.61	6	2	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	3.001805	-73.94306	0.37	0.47	0.47	5	0	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80739	-73.94311	0.59	0.69	0.69	6	3	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80739	-73.94311	0.14	0.14	0.14	2	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80739	-73.94309	0.2	0.22	0.22	7	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80738	-73.94312	0.2	0.22	0.22	7	0	0	0	0	0	0	0	0	0	0	0	0																											
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350030000	158-E9	40.80738	-73.94313	0.31	0.85	0.89	8	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80738	-73.94316	0.56	0.65	0.65	10	1	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80736	-73.94312	0.21	0.22	0.22	7	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80735	-73.94312	0.42	0.44	0.44	9	2	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80735	-73.94312	0.56	0.99	1.01	3	3	0	1	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.42	0.44	0.44	9	2	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.18	0.19	0.19	6	0	0	0	0	0	0	0	0	0	0	0	0																											
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350030000	158-E9	40.80733	-73.94312	0.54	1.67	1.83	5	2	0	0	1	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.54	0.63	0.64	7	2	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.24	0.24	0.24	8	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.43	0.45	0.45	7	3	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.21	0.22	0.22	8	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.15	0.16	0.16	5	0	0	0	0	0	0	0	0	0	0	0	0																											
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350030000	158-E9	40.80733	-73.94312	0.49	0.51	0.51	11	2	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.41	0.5	0.5	5	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80733	-73.94312	0.7	0.87	0.87	9	1	2	0	0	0	0	0	0	0	0	0	0																											
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350030000	158-E9	40.80734	-73.94312	0.38	1.69	1.69	3	2	0	0	0	1	0	0	0	0	0	0	0																											
350030000	158-E9	40.80734	-73.94312	0.15	0.16	0.16	6	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80734	-73.94312	0.29	0.3	0.3	7	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.8074	-73.94312	0.5	0.59	0.6	6	2	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.8074	-73.94312	0.31	0.32	0.32	8	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80742	-73.94312	0.44	0.55	0.55	5	2	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80742	-73.94312	0.49	0.51	0.51	7	4	0	0	0	0	0	0	0	0	0	0	0																											
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350030000	158-E9	40.80743	-73.94312	0.28	0.29	0.29	7	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80743	-73.94312	0.64	3.28	38.26	11	2	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80743	-73.94312	0.45	0.47	0.47	7	3	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80748	-73.94312	0.26	0.27	0.27	9	0	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80748	-73.94312	0.59	0.69	0.69	4	4	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80748	-73.94312	0.22	0.23	0.23	5	1	0	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80741	-73.94312	0.45	0.55	0.55	7	1	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80741	-73.94312	0.29	0.38	0.38	4	0	1	0	0	0	0	0	0	0	0	0	0																											
350030000	158-E9	40.80741	-73.94																																											



Data Description -- Common

deviceID: "1f004a000d504e5354303
420"

time: 1631239374

latitude: 40.84767

longitude: -73.8693

deviceID:

- Unique identifier for each City Scanner Device

Time:

- Epoch time
- # of seconds that have elapsed since January 1 1970 (midnight UTC/GMT) not counting leap seconds

Latitude/Longitude:

- Unit: Degrees

Data Description -- raw data

```
bin0:5945      PM1:1.63
bin1:418       PM25:3.81
bin2:101        PM10:21.58
...
bin23:0         gas_op2_w:654
                  temperature:23.5
                  humidity:71.9
                  noise:32
```

24 Bins:

- Separate particle count by size
- Unit: # (count)

PM1:

- Particulate matter ~1 micron in diameter
- Units: ug/m³

PM2.5

- Particulate matter ~2.5 microns in diameter
- Units: ug/m³

PM10:

- Particulate matter ~10 microns in diameter
- Units: ug/m³

Gas_op2_w

- Electric signal for NO₂
- Units: mv

temperature

- Ambient temperature
- Units: Degrees celsius

humidity

- Ambient humidity
- Units: % Relative humidity out of 100%

noise:

- Units: Voltage level in mV

Data Description -- Calibrated NO2

tmpf: 20

dwpf: 12.78

relh: 63.12

drct: 310

sknt: 7.20

mslp: 101.1

vsby: 16.1

feel: 20

Calib_logNO2: 2.43

Calib_NO2: 11.32

Spline_10min: 11.22

Spline_dmean: 4.75

Bckadj_NO2: 4.85

tmpf:

- Temperature at nearest weather station
- Units: Degrees celsius

dwpf:

- Dewpoint at nearest weather station
- Units: Degrees celsius

relh:

- Relative humidity at nearest weather station
- Units: %

drct:

- Wind direction with reference to the true north as 0
- Units: Degrees

sknt:

- Wind speed at nearest weather station
- Units: m/s

mslp:

- Air pressure at nearest weather station
- Units: kpa

vsby:

- Visibility at nearest weather station
- Units: km

feel:

- Feel like temperature at nearest weather station
- Units: Degrees celsius

Calib_logNO2:

- Calibrated NO2 in log form
- Units: log ppb

Calib_NO2

- Calibrated NO2
- Units: ppb

Spline_10min

- Spline regressed NO2 using 10 min minimum values
- Units: ppb

Spline_dmean

- Daily median NO2
- Units: ppb

Bckadj_NO2

- Background adjusted NO2 after calibration
- Units: ppb

Data Description -- Calibrated PM2.5

tmpf: 20

dwpf: 12.78

relh: 63.12

drct: 310

sknt: 7.20

mslp: 101.1

vsby: 16.1

feel: 20

Calib_logPM: 1.34

Calib_PM: 3.82

Spline_10min: 4.37

Spline_dmean: 4.02

Bckadj_PM: 3.47

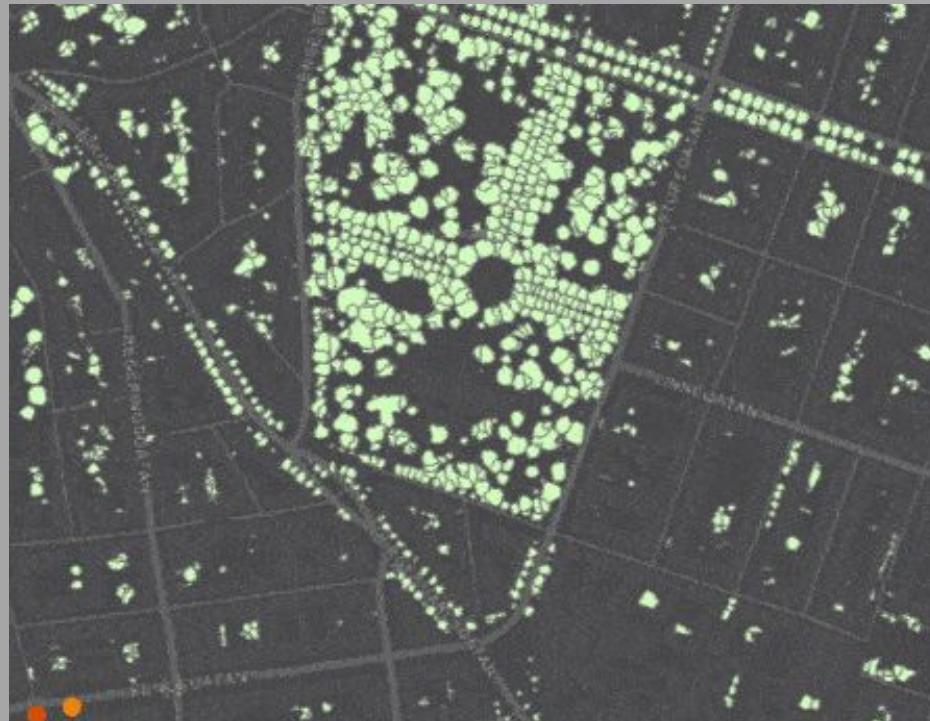
- tmpf:**
 - Temperature at nearest weather station
 - Units: Degrees celsius
- dwpf:**
 - Dewpoint at nearest weather station
 - Units: Degrees celsius
- relh:**
 - Relative humidity at nearest weather station
 - Units: %
- drct:**
 - Wind direction with reference to the true north as 0
 - Units: Degrees
- sknt:**
 - Wind speed at nearest weather station
 - Units: m/s
- mslp:**
 - Air pressure at nearest weather station
 - Units: kpa
- vsby:**
 - Visibility at nearest weather station
 - Units: km
- feel:**
 - Feel like temperature at nearest weather station
 - Units: Degrees celsius
- Calib_logPM:**
 - Calibrated PM in log form
 - Units: log ug/m³
- Calib_PM**
 - Calibrated PM
 - Units: ug/m³
- Spline_10min**
 - Spline regressed PM using 10 min minimum values
 - Units: ug/m³
- Spline_dmean**
 - Daily median PM
 - Units: ug/m³
- Bckadj_PM**
 - Background adjusted PM after calibration
 - Units: ug/m³

Activity: Methodology

Understanding Spatial Environmental Patterns

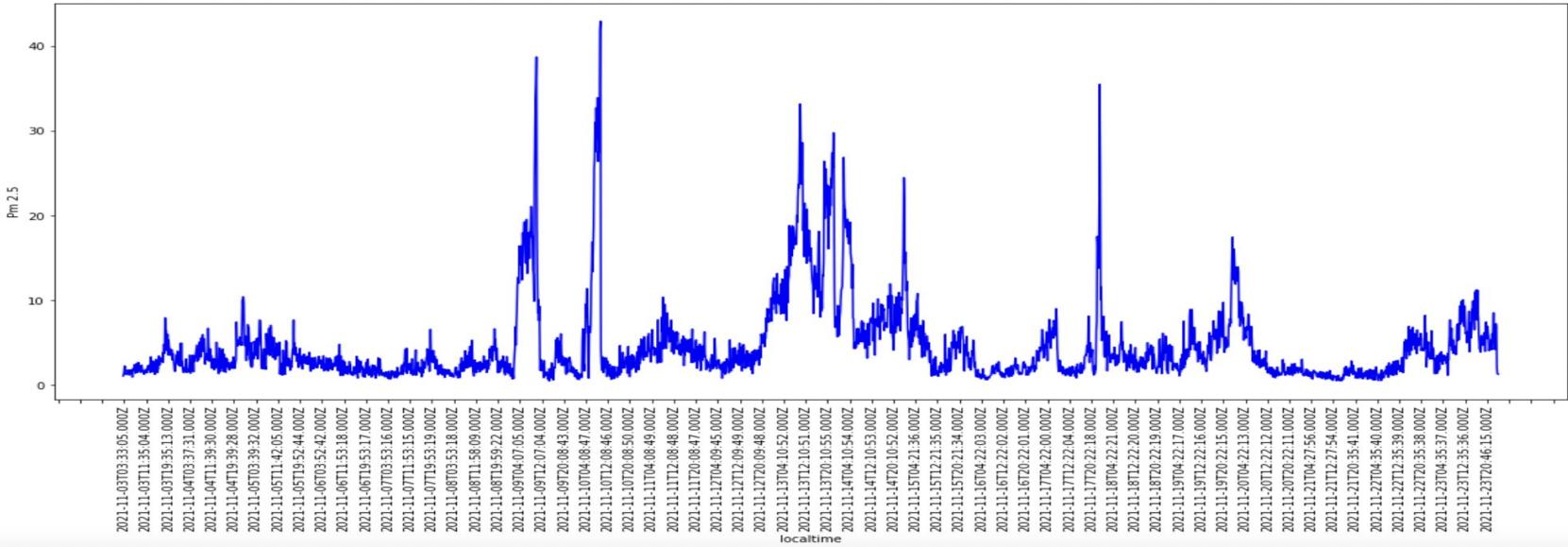
DATA ANALYSIS METHODS

How do we understand the insights this data can provide?
Overview of key CityScanner analysis practices



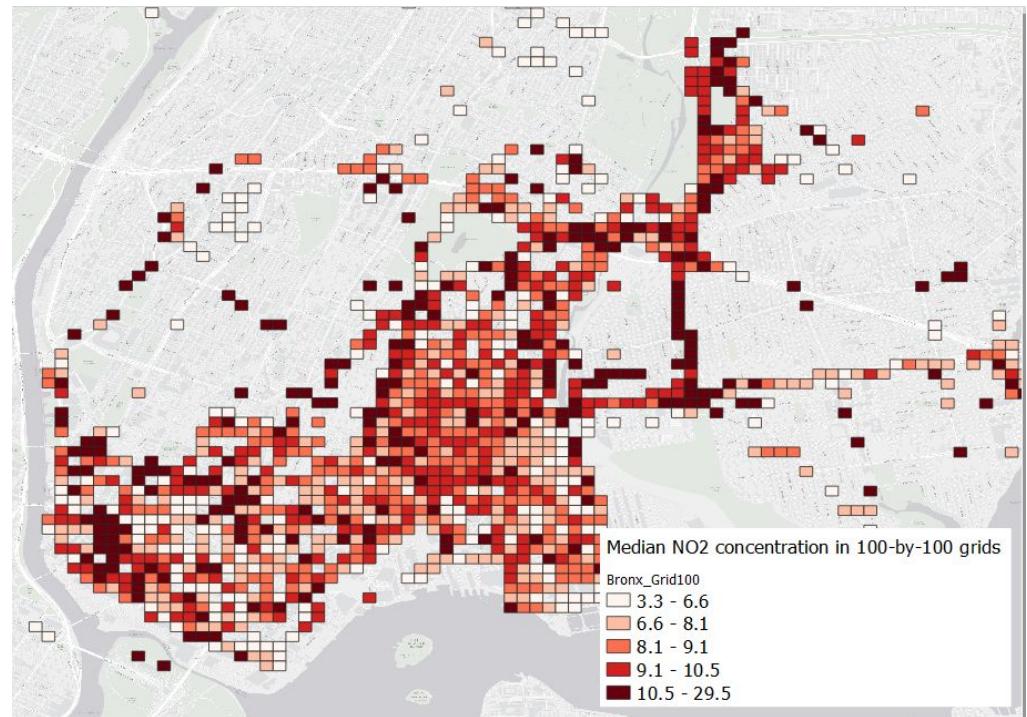
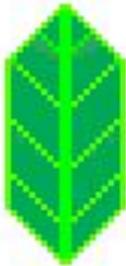
Analysis Method: Time Series

- Parameter as a function of time



Analysis Method: Mapping

- What can maps tell us?
- What info do you need to make a map?
- What tools can you use for mapping?



Analysis Method: Hotspot Detection

"A hot-spot analysis is defined in 40 CFR 93.101 as an estimation of likely future localized pollutant concentrations and a comparison of those concentrations to the relevant NAAQS." - United States EPA

Analysis Method: Hotspot Detection

Clustering

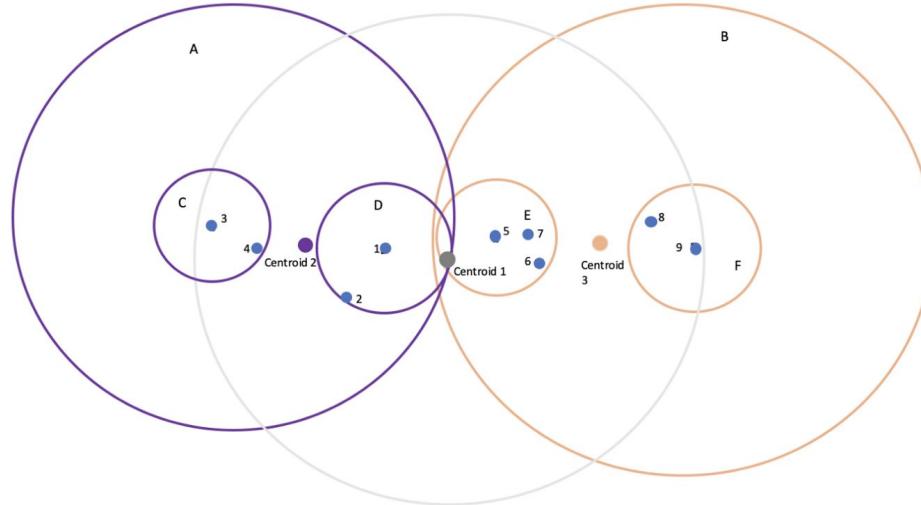
- DBSCAN (from scikit learn)
 - Density-Based Spatial Clustering of Applications with Noise
 - Finds core samples of high density and expands clusters from them
 - Good for data which contains clusters of similar density.



Analysis Method: Hotspot Detection

Clustering

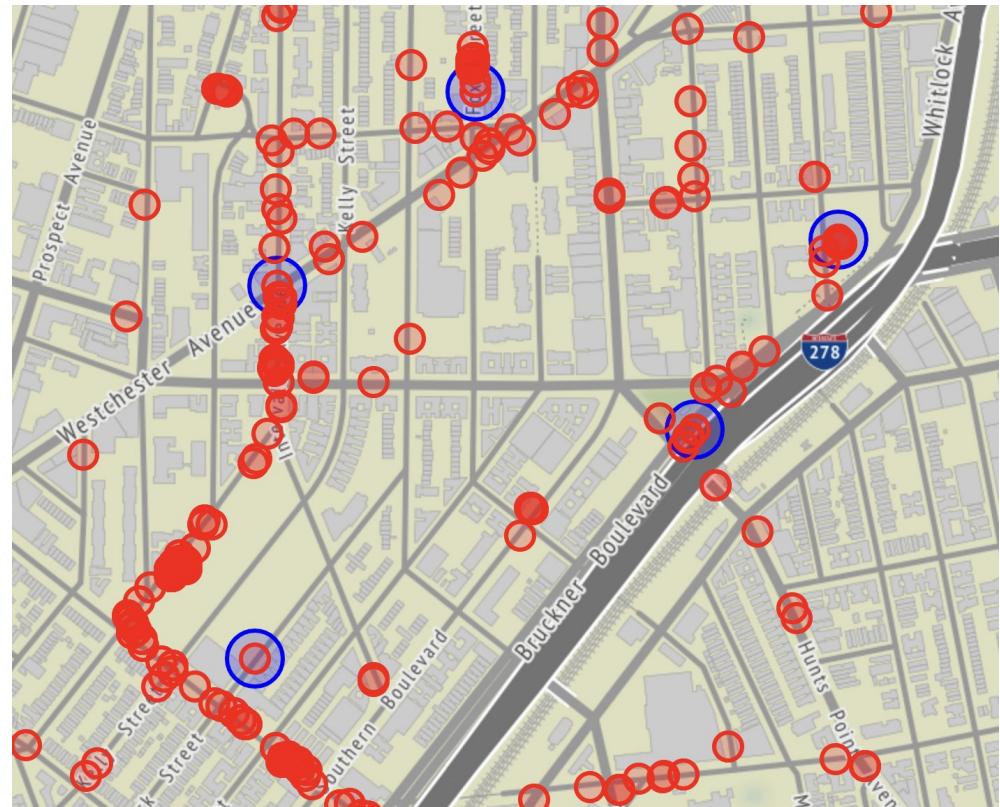
- DBSCAN → Ball_tree algorithm
 - Divides groups of points into clusters until desired size is reached



Analysis Method: Hotspot Detection

Hotspot Detection

- Bottom-up hierarchical clustering
“agglomerative”
- Merge clusters of data into smaller clusters



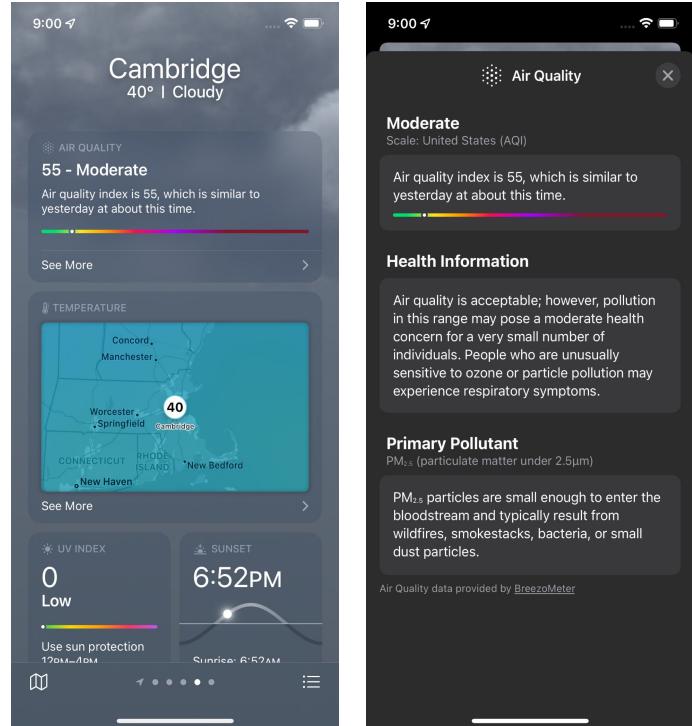
Air Quality API Exercise

World Air Quality Index API

Air Quality Index

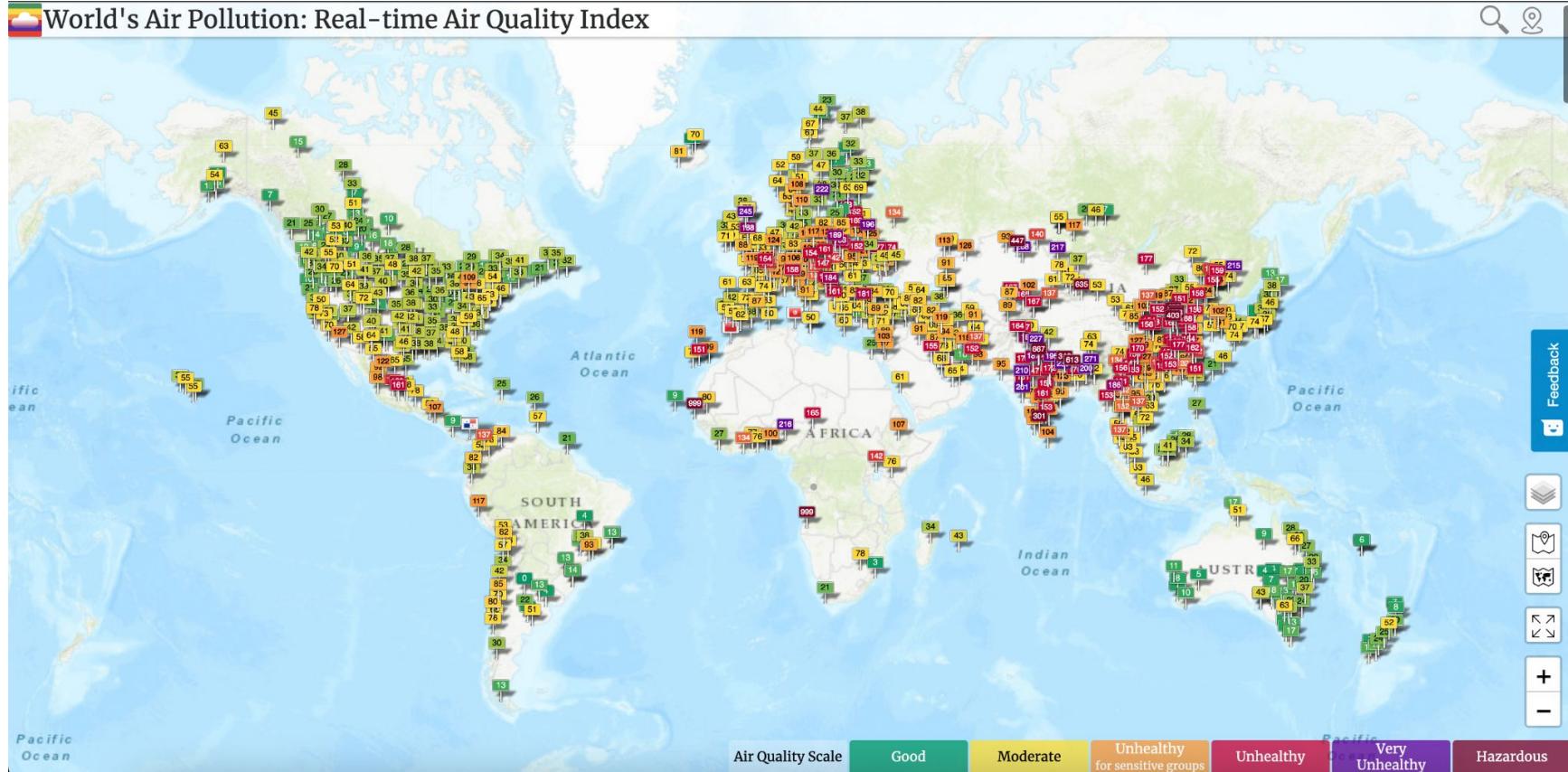
Air Quality Index		
AQI Category and Color	Index Value	Description of Air Quality
Good Green	0 to 50	Air quality is satisfactory, and air pollution poses little or no risk.
Moderate Yellow	51 to 100	Air quality is acceptable. However, there may be a risk for some people, particularly those who are unusually sensitive to air pollution.
Unhealthy for Sensitive Groups Orange	101 to 150	Members of sensitive groups may experience health effects. The general public is less likely to be affected.
Unhealthy Red	151 to 200	Some members of the general public may experience health effects; members of sensitive groups may experience more serious health effects.
Very Unhealthy Purple	201 to 300	Health alert: The risk of health effects is increased for everyone.
Hazardous Maroon	301 and higher	Health warning of emergency conditions: everyone is more likely to be affected.

Air quality index ([United States Environmental Protection Agency](#))



Air quality index shown in iPhone weather app

Accessing (Global) EPA Data



Accessing (Global) EPA Data

WAQI

- Site Link: <https://waqi.info/>
- API Link: <https://aqicn.org/data-platform/token/>

Get API Token:

- <https://aqicn.org/data-platform/token/>

Notebook Setup:

- Python libraries

Authentication

- Input custom token into notebook

Getting the Data

- Run sample code
- Modify + experiment!

Discussion

Understanding Spatial Environmental
Patterns