

Cirrolytix Research Services

PROJECT AEDES ENHANCEMENT

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Digital
Public
Goods
Alliance
unicef 

Solutions and Enhancements

PROJECT AEDES OPEN PLATFORM

FORECASTING

Predict future number of cases/deaths of vector-borne diseases

HOTSPOT DETECTION

Identify locations of possible hotspots for outbreaks

INFORM RISK MAPPING

Map out risk framework using environmental data

OPEN API

Publicly open pre-processed satellite, weather, socioeconomic and health datasets

PYTHON PACKAGE

Open-source tools used for data collection, feature engineering and automated machine learning

RISK-BASED ASSESSMENT FRAMEWORK

HAZARDS

Monitor progress
of epidemic,
Generate alerts

VULNERABILITIES

Prioritize areas
with vulnerable
groups, suggest
demographic and
geographic
determinants of
risk

LACK OF COPING CAPACITY

Prioritize areas for
emergency aid,
recommend
infrastructure
investment

IDEAL RISK-BASED ASSESSMENT FRAMEWORK

HAZARDS

Dengue Case incidence
Flood Occurrence
Temperature
Precipitation
COVID-19 Incidence
Access to water
Access to sanitation

VULNERABILITIES

Population ages 0-20
Poverty Index
Population affected by natural disasters
Population previously infected by dengue
Mortality
Land-use types
Social listening
Primary and secondary schools
PhilHealth coverage
Human mobility

LACK OF COPING CAPACITY

Presence of health centers
Presence of hospitals
Number of health workers
Health expenditure
Vaccination coverage

DPG RISK-BASED ASSESSMENT FRAMEWORK

HAZARDS

Dengue Case
Temperature
Precipitation Rate
Relative Humidity
Water Index
Distance and count of
water sources
Distance and count of
sanitation and waste
facilities

VULNERABILITIES

Vegetation Index
Built-up Index
Soil Moisture Index
Google search trends
Distance and count of
kindergartens
Distance and count of schools
Distance and count of colleges
Distance and count of
universities

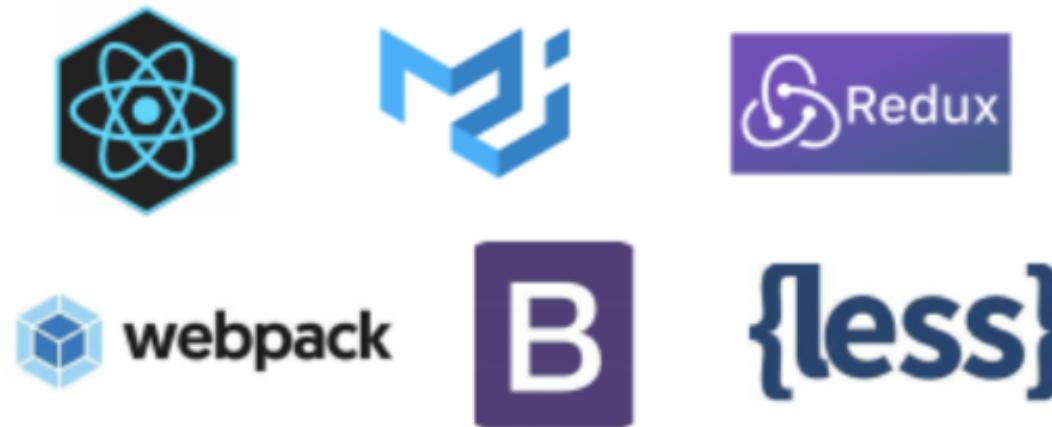
LACK OF COPING CAPACITY

Distance and count of clinics
Distance and count of
hospitals
Distance and count of doctors

FRONT-END | BASICS



FRONT-END | FRAMEWORKS



FRONT-END | TOOLS



 Data-Driven Documents

BACK-END



DATABASE | RDBMS



DEV-OPS | INFRASTRUCTURES



ENHANCED TECH STACK

DEV-OPS | DEVELOP

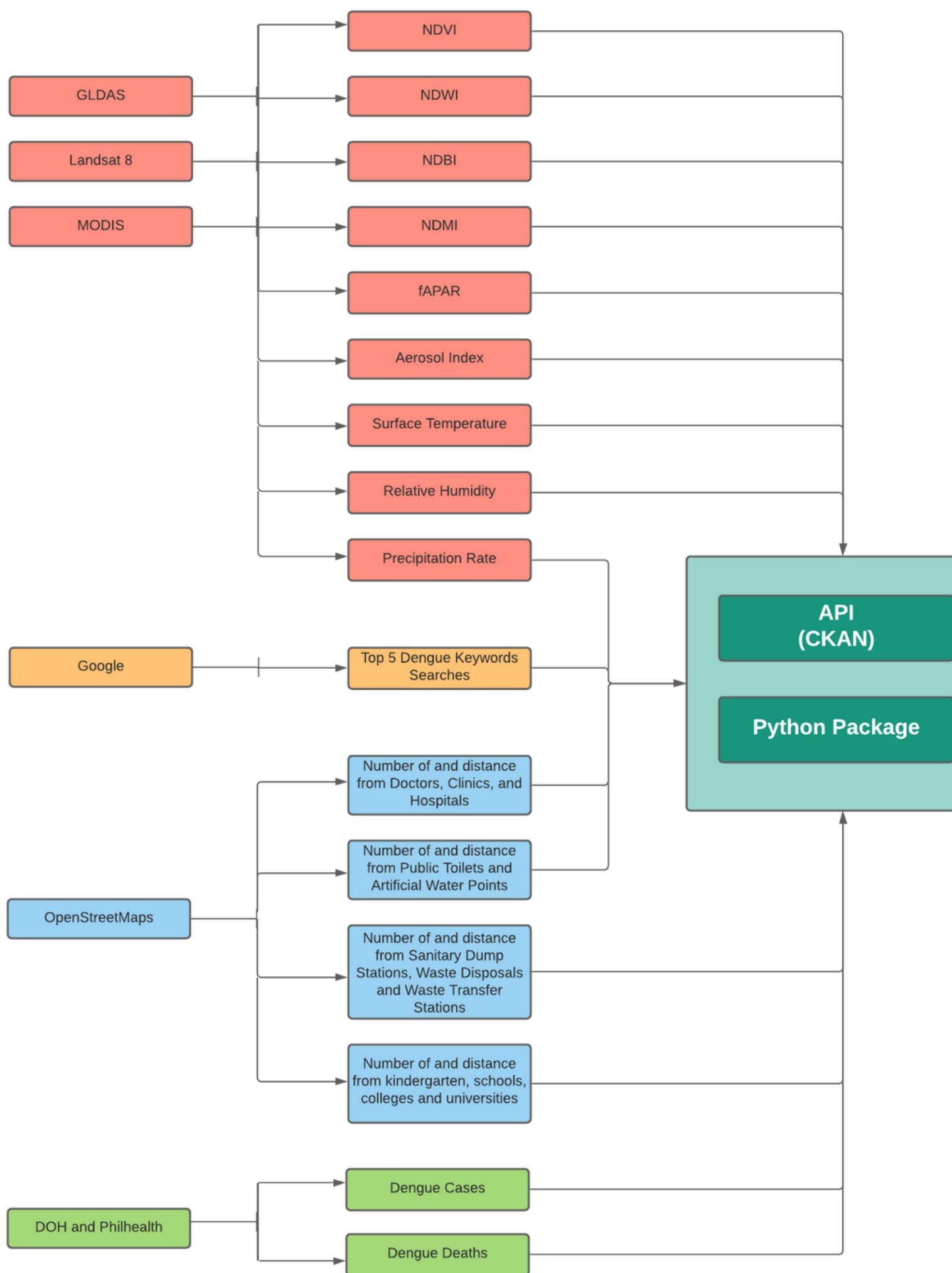


DATA SERVICES



DATA MANAGEMENT SYSTEM

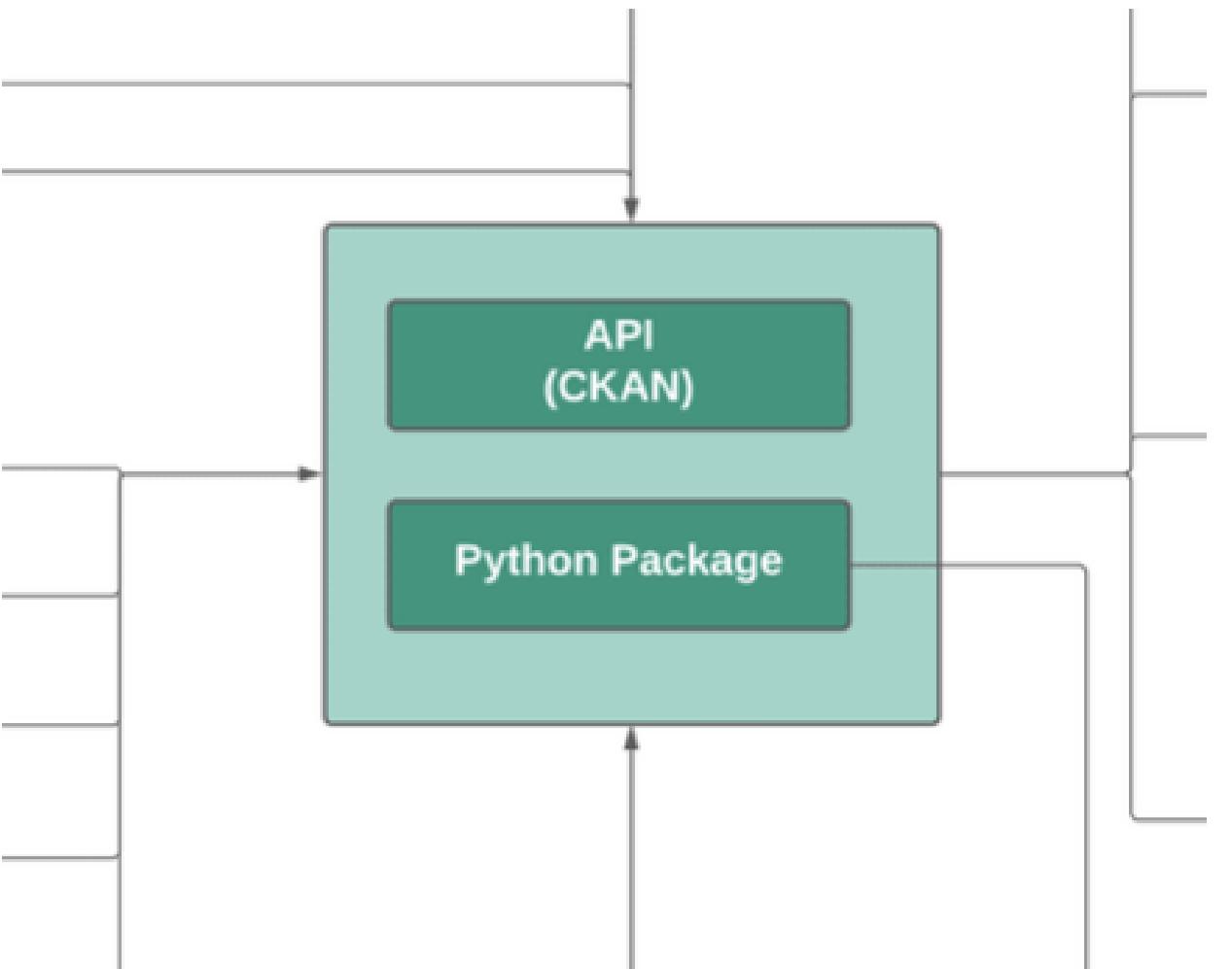




Enhancement Data Model

Data Management System

Python
Package



**INSERT SLIDE ON
BEFORE/AFTER FOR QGIS vs
AEDES PYTHON PACKAGE**

**INSERT SLIDE ON PROBLEMS
BEING SOLVED FROM PREVIOUS
ITERATION (NASA HACKATHON
VS ENHANCEMENTS)**

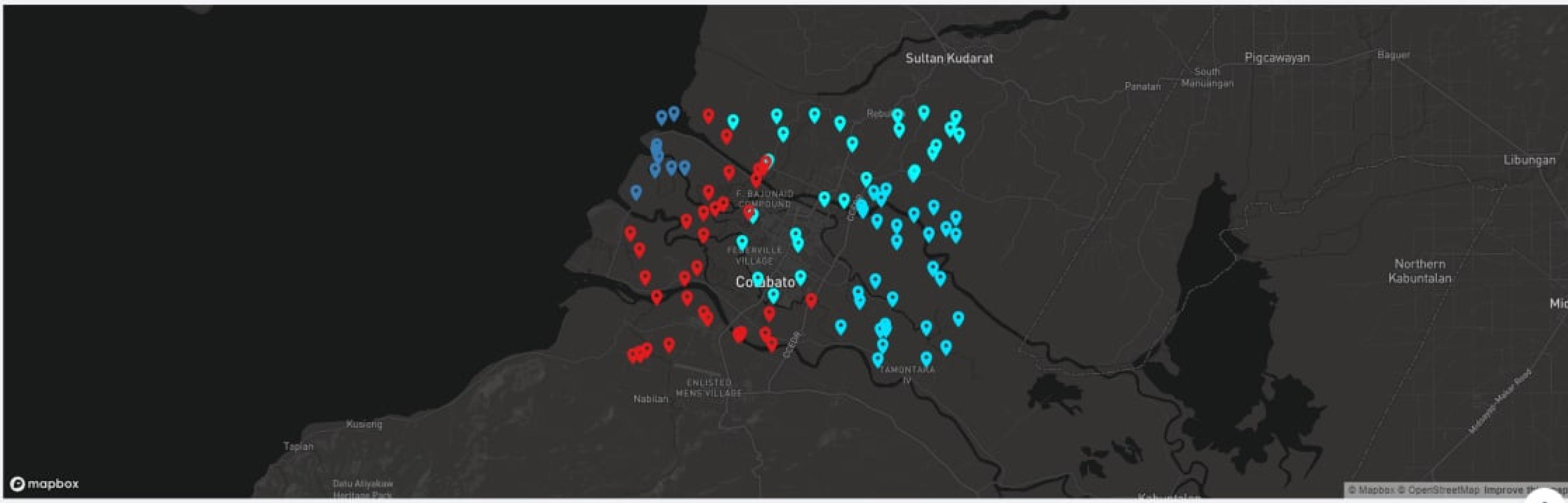
Dashboard

Points with Remote Sensing

ID	LONGITUDE	LATITUDE	NDVI	NDWI	NDMI	NDVI	NDWI	NDMI	NAME	ACROSSL	LABELS	INDEX	PLACE ID	LICENCE	DATA TO
1	124.23271102909972	7.2322490886052195	0.29426543772184005	-0.18135909206120807	-0.10210229297665244	137.24185463659148	0	0	S211140494	"Data © OpenStreetMap contributors	00bL				
2	124.188791024222	7.1651933675878565	0.33712953424396913	-0.17811715393669746	-0.1372320948832157	210.6881987577764	0	0	S242136719	"Data © OpenStreetMap contributors	00bL				
3	124.20985630039459	7.196758407621468	0.26254398035434753	-0.16634924621562536	-0.07652650886410261	148.53673723536738	0	0	S065253603	"Data © OpenStreetMap contributors	00bL				
4	124.19676303625263	7.2513988770380315	0.049508159867661314	-0.08317019477819493	0.04097183914503763	190.84893882646682	1	0	S248277417	"Data © OpenStreetMap contributors	00bL				
5	124.29736612018179	7.218621620798705	0.24762520238554964	-0.1811293855035335	-0.04423150200908271	171.52970922882432	2	0	S665139777	"Data © OpenStreetMap contributors	00bL				
6	124.30661530426687	7.178069029871818	0.3625078864123668	-0.2336097880415473	-0.09924217118091	168.4800745527	2	0	S156615333	"Data © OpenStreetMap contributors	00bL				
7	124.26744081993178	7.2416938171902725	0.3057146163256264	-0.1561891062137156	-0.1371504720069124	165.52970922882432	3	0	S665615333	"Data © OpenStreetMap contributors	00bL				
8	124.29377210327505	7.253233563803932	0.3395442103023234	-0.18009947910026575	-0.1443210839604	150.52970922882432	4	0	S130815308	"Data © OpenStreetMap contributors	00bL				
9	124.2778049714143	7.173807474207336	0.3199756208166676	-0.2031017028784548	-0.0931094008004132	151.52970922882432	5	0	S188621063	"Data © OpenStreetMap contributors	00bL				
10	124.26291924257728	7.249400110845375	0.3182352310440857	-0.16604603124972356	-0.1403106200755383	152.52970922882432	6	0	S188621063	"Data © OpenStreetMap contributors	00bL				

Open Data through API and Dashboard Design

<https://aedes-datacatalogue-beta.xyz/>





Visualize on a Map

This package also provides the capability of visualizing all the points on a map. [PyPI: https://pypi.org/project/aedes/](https://pypi.org/project/aedes/)
repository: <https://github.com/xmpuspus/aedes>

```
vizo = visualize_on_map(rev_geocode_qc_df)  
vizo
```



aedes

A python package for PROJECT AEDES by Xavier Puspus of Cirrolytix Research Services.



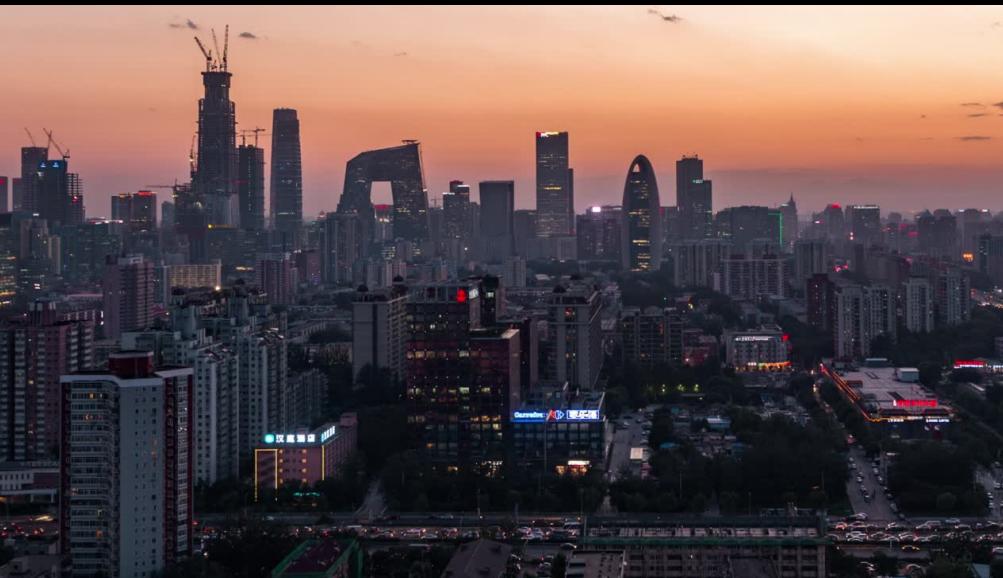
Dengue Detection Features

COLLECTING SATELLITE DATA
AT POINTS OF INTEREST



Vegetation and Water

NDWI
NDVI
FAPAR
NDMI



Urbanicity

NDBI
Aerosol Index



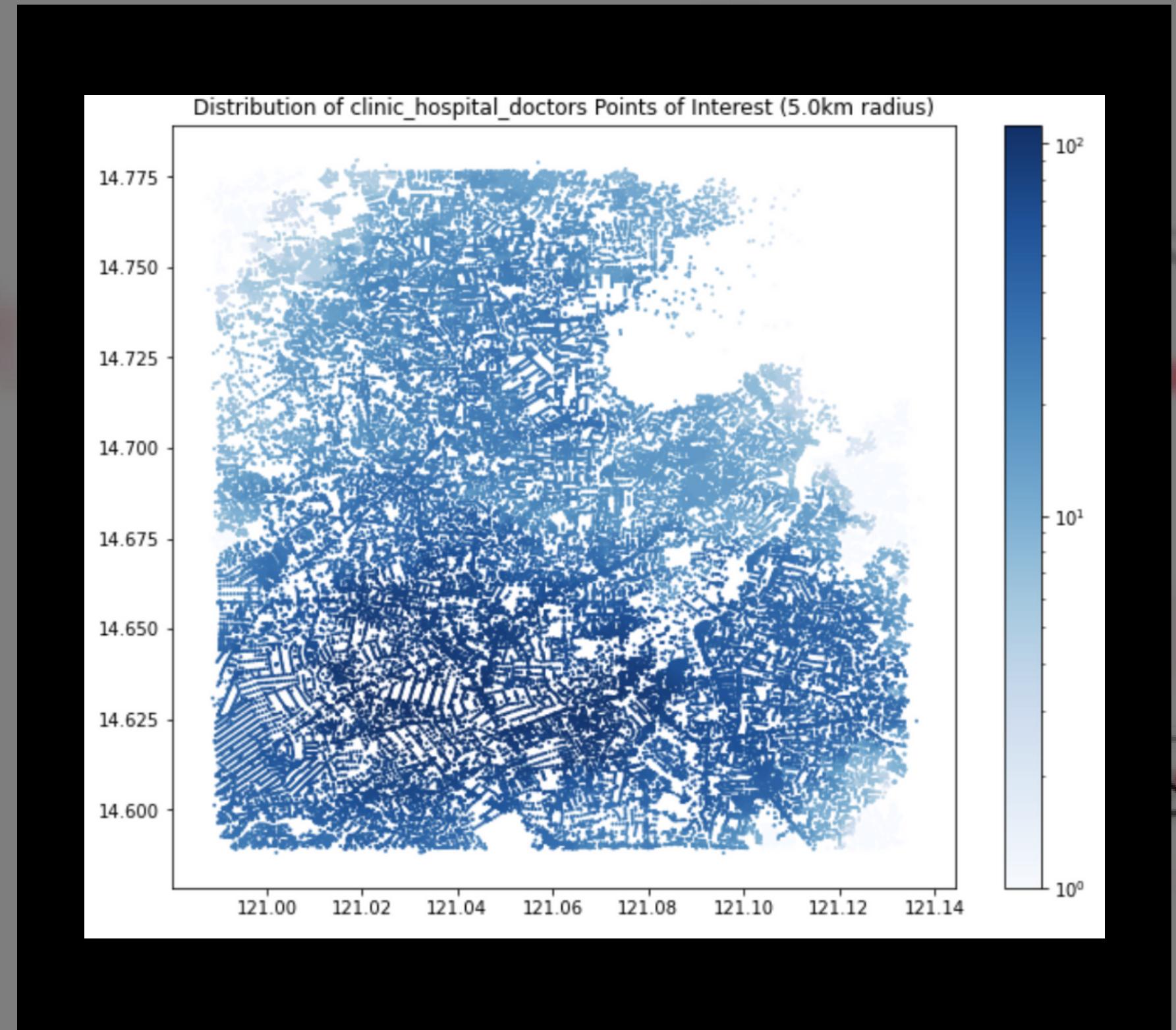
Weather

Surface
Temperature
Precipitation Rate
Relative Humidity

Healthcare Capacity

DISTRIBUTION PER RADIAL BLOCK

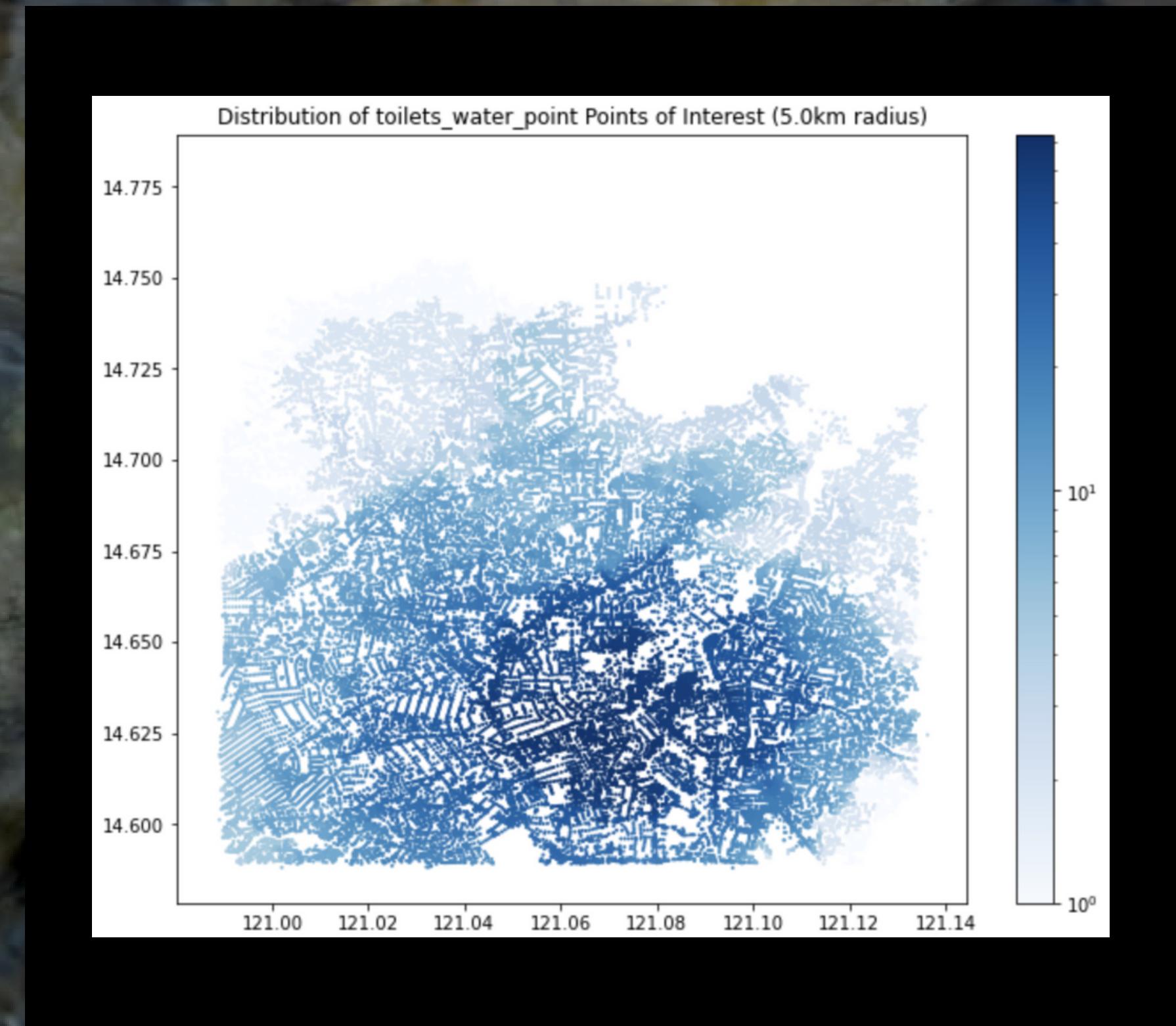
Heatmap of **accessibility** to nearest clinic, and hospital



Access to Water Sources

DISTRIBUTION PER RADIAL
BLOCK

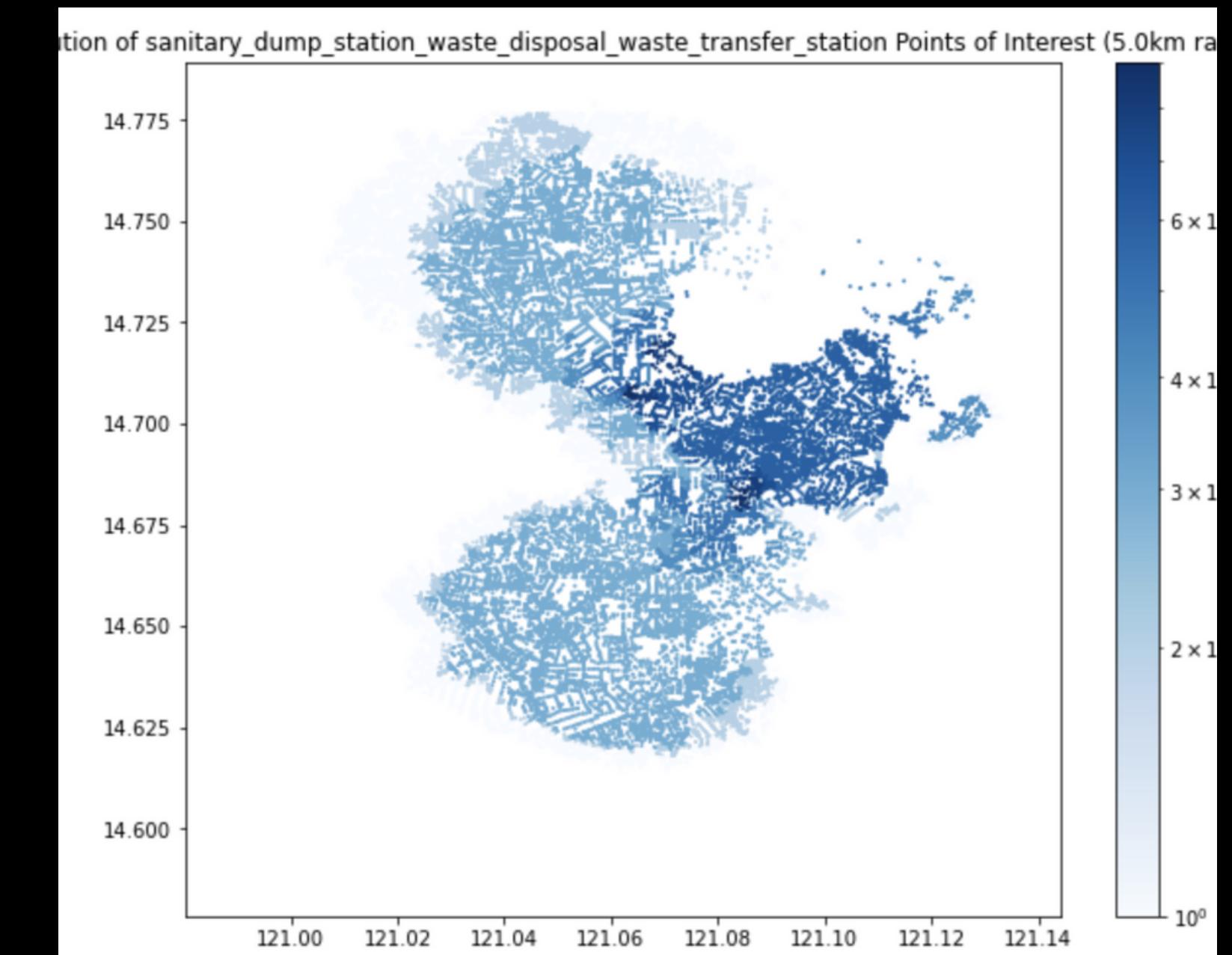
Heatmap of **accessibility** to nearest
toilet facility and different types of
water points as defined by OSM



Access to Sanitation

DISTRIBUTION PER RADIAL BLOCK

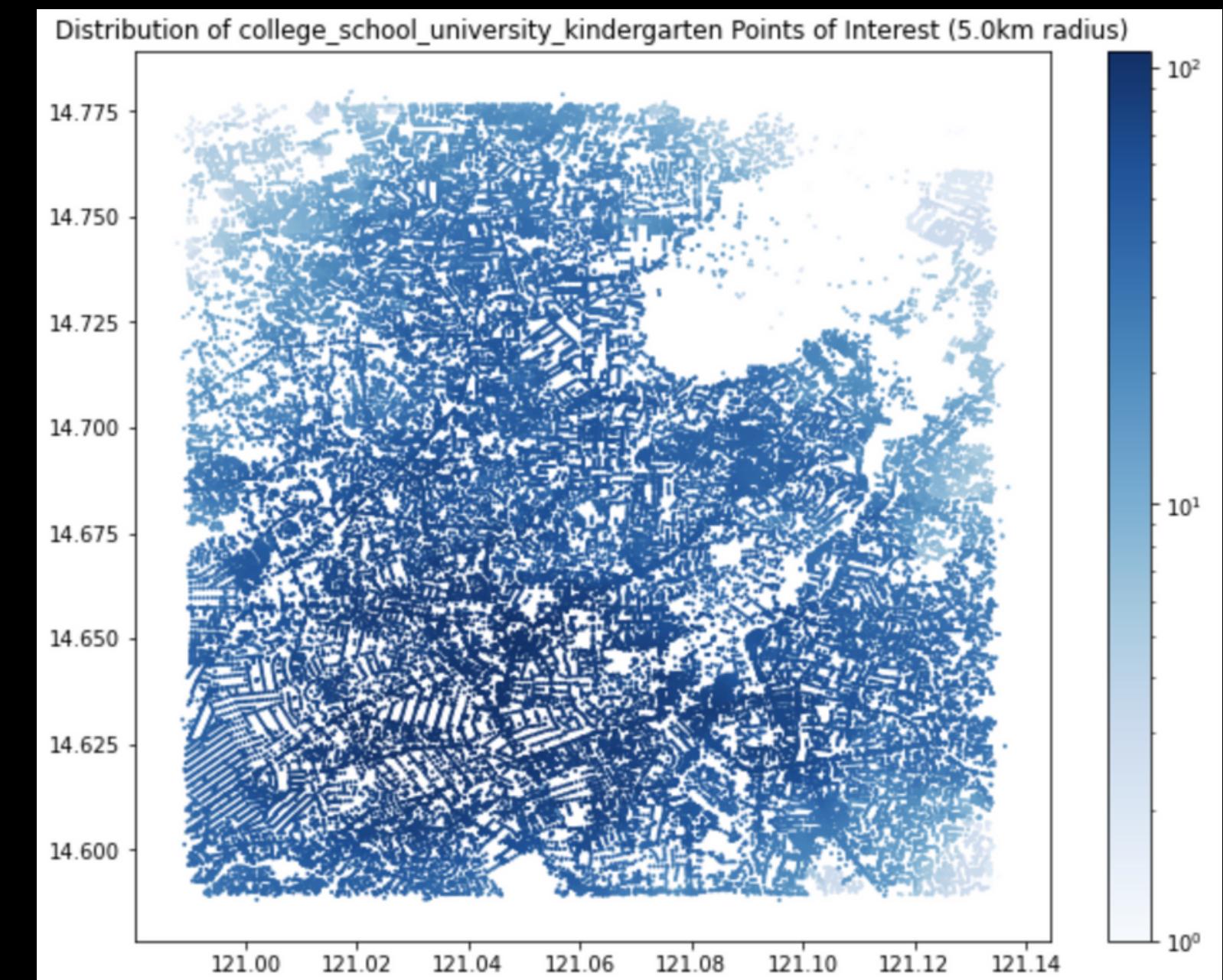
Heatmap of **accessibility** to nearest **sanitary dump station, waste disposal, and waste transfer stations**



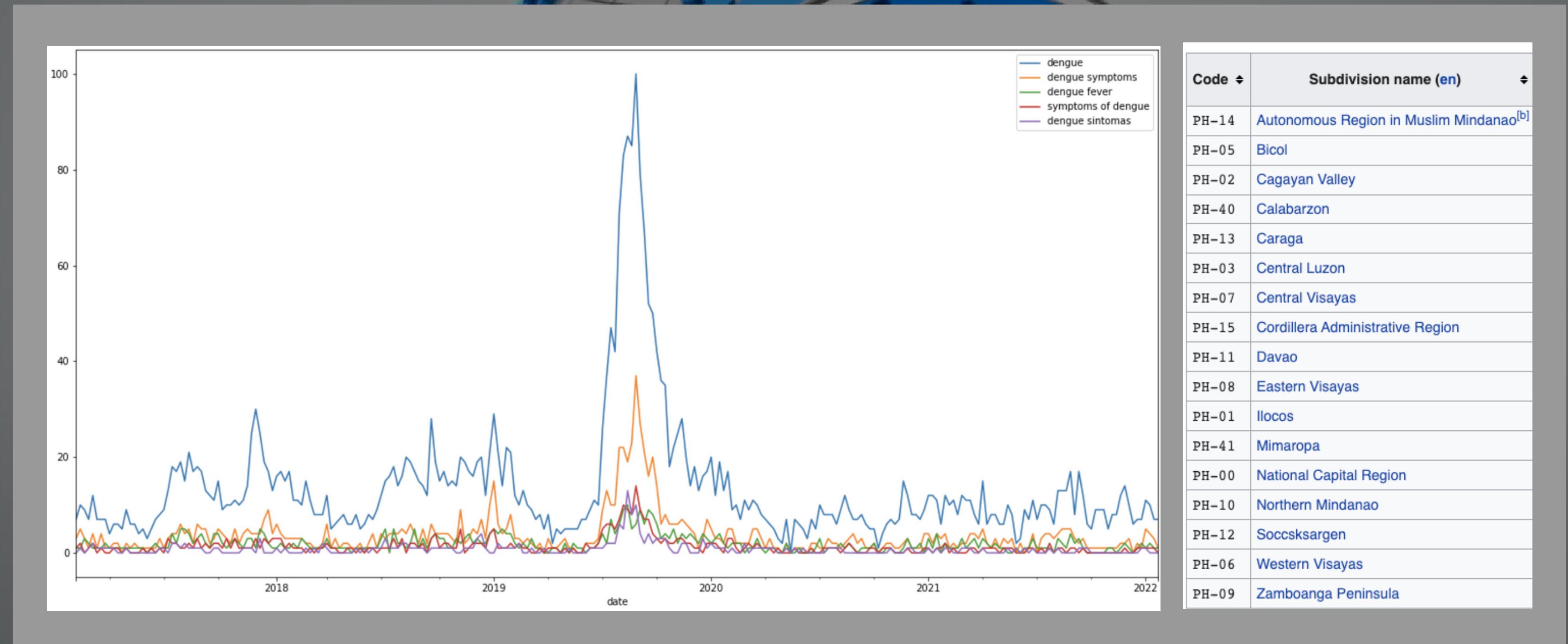
Schools

DISTRIBUTION PER RADIAL BLOCK

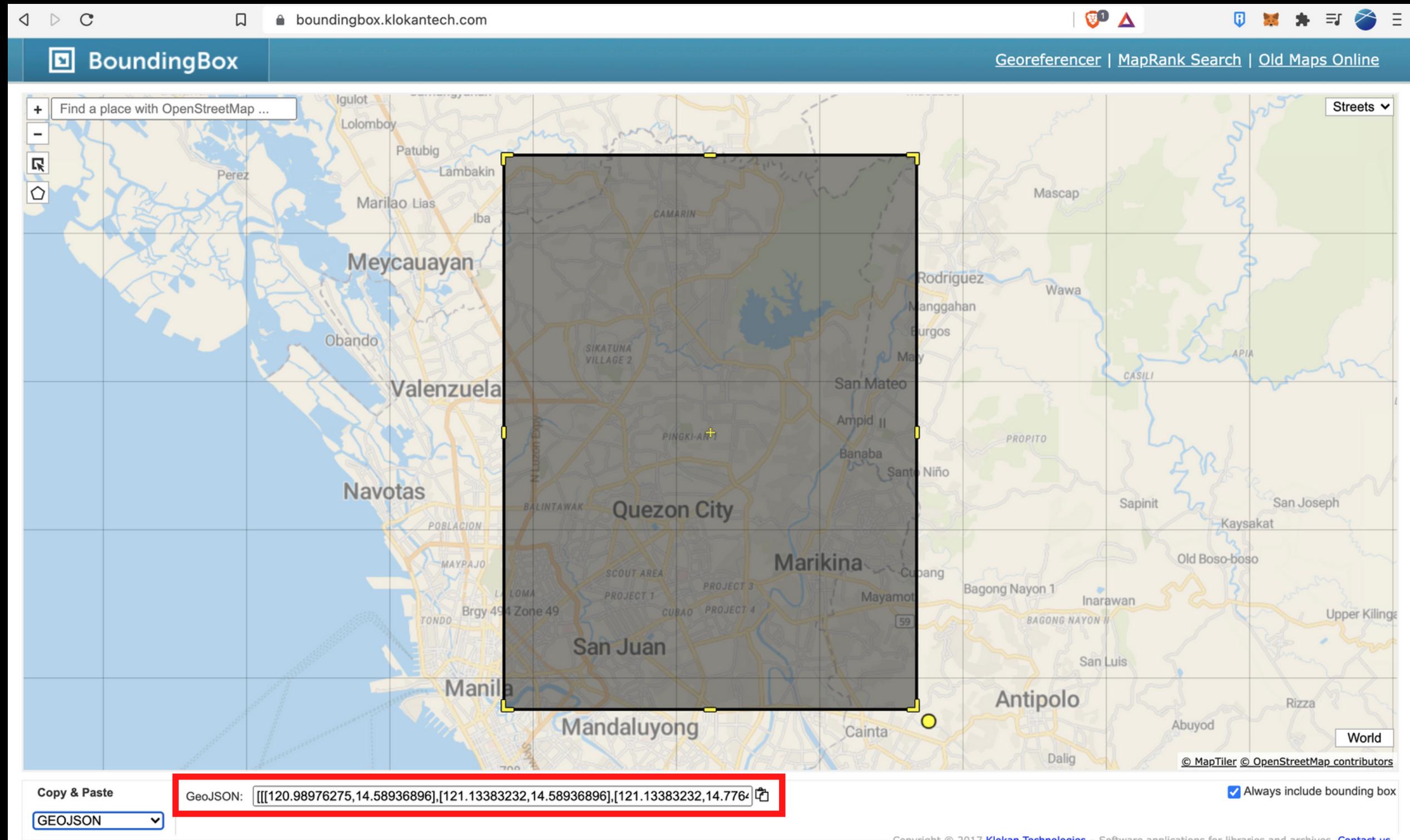
Heatmap of **accessibility to nearest kindergarten, schools, colleges, and universities**



Google Search Trends



Sample Web App Data Input



In order to collect the data and features necessary for modelling, the only input is a **bounding box geojson or a polygon**

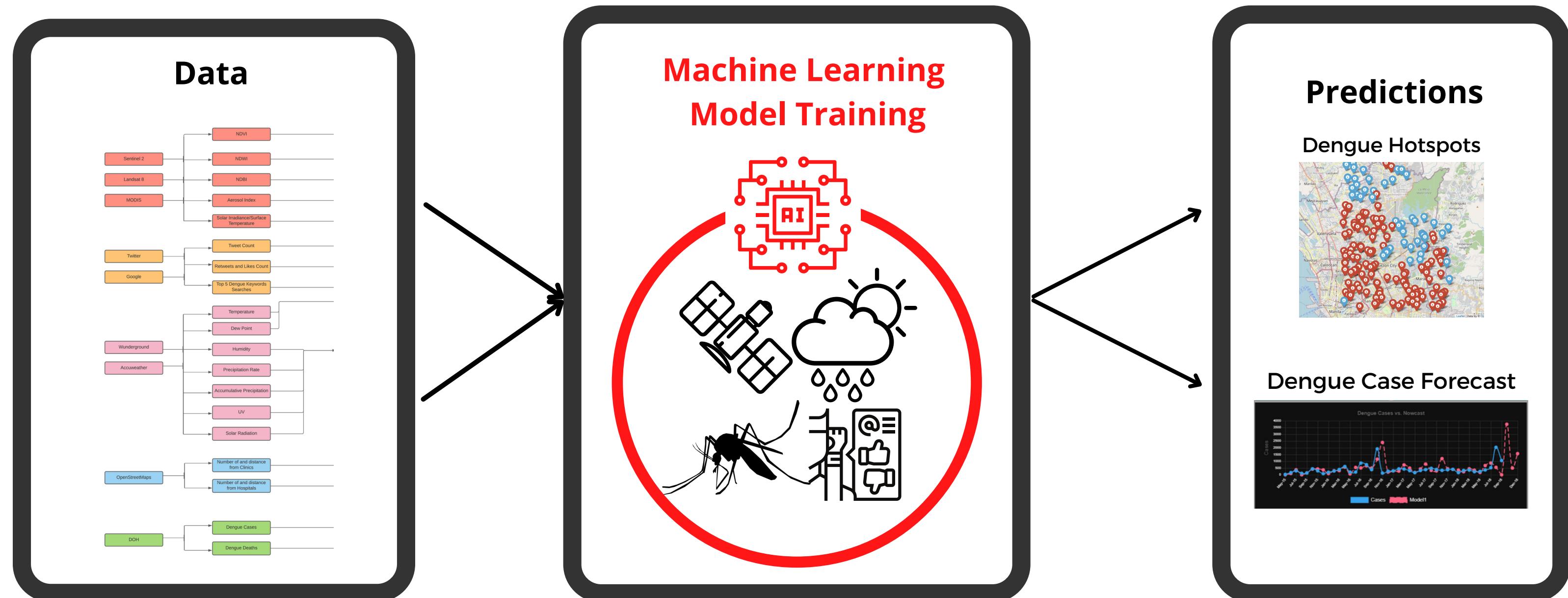
Extracted Data Features

#	Column	Non-Null Count	Dtype
0	geometry	50 non-null	geometry
1	buffered_geometry	50 non-null	object
2	longitude	50 non-null	float64
3	latitude	50 non-null	float64
4	ndvi	50 non-null	float64
5	fapar	50 non-null	float64
6	ndbi	50 non-null	float64
7	ndwi	50 non-null	float64
8	ndmi	50 non-null	float64
9	aerosol	50 non-null	float64
10	surface_temperature	50 non-null	float64
11	precipitation_rate	50 non-null	float64
12	relative_humidity	50 non-null	float64
13	labels	50 non-null	int32
14	OSM_network_id	50 non-null	int64
15	nearest_clinic_hospital_doctors_1	50 non-null	float64
16	nearest_clinic_hospital_doctors_2	50 non-null	float64
17	nearest_clinic_hospital_doctors_3	50 non-null	float64
18	nearest_clinic_hospital_doctors_4	50 non-null	float64
19	nearest_clinic_hospital_doctors_5	50 non-null	float64
20	count_clinic_hospital_doctors_within_5.0km	50 non-null	float64
21	nearest_toilets_water_point_1	50 non-null	float64
22	nearest_toilets_water_point_2	50 non-null	float64
23	nearest_toilets_water_point_3	50 non-null	float64
24	nearest_toilets_water_point_4	50 non-null	float64
25	nearest_toilets_water_point_5	50 non-null	float64
26	count_toilets_water_point_within_5.0km	50 non-null	float64
27	nearest_sanitary_dump_station_waste_disposal_waste_transfer_station_1	50 non-null	float64
28	nearest_sanitary_dump_station_waste_disposal_waste_transfer_station_2	50 non-null	float64
29	nearest_sanitary_dump_station_waste_disposal_waste_transfer_station_3	50 non-null	float64
30	nearest_sanitary_dump_station_waste_disposal_waste_transfer_station_4	50 non-null	float64
31	nearest_sanitary_dump_station_waste_disposal_waste_transfer_station_5	50 non-null	float64
32	count_sanitary_dump_station_waste_disposal_waste_transfer_station_within_5.0km	50 non-null	float64
33	nearest_college_school_university_kindergarten_1	50 non-null	float64
34	nearest_college_school_university_kindergarten_2	50 non-null	float64
35	nearest_college_school_university_kindergarten_3	50 non-null	float64
36	nearest_college_school_university_kindergarten_4	50 non-null	float64
37	nearest_college_school_university_kindergarten_5	50 non-null	float64
38	count_college_school_university_kindergarten_within_5.0km	50 non-null	float64

dtypes: float64(35), geometry(1), int32(1), int64(1), object(1)

ML Model

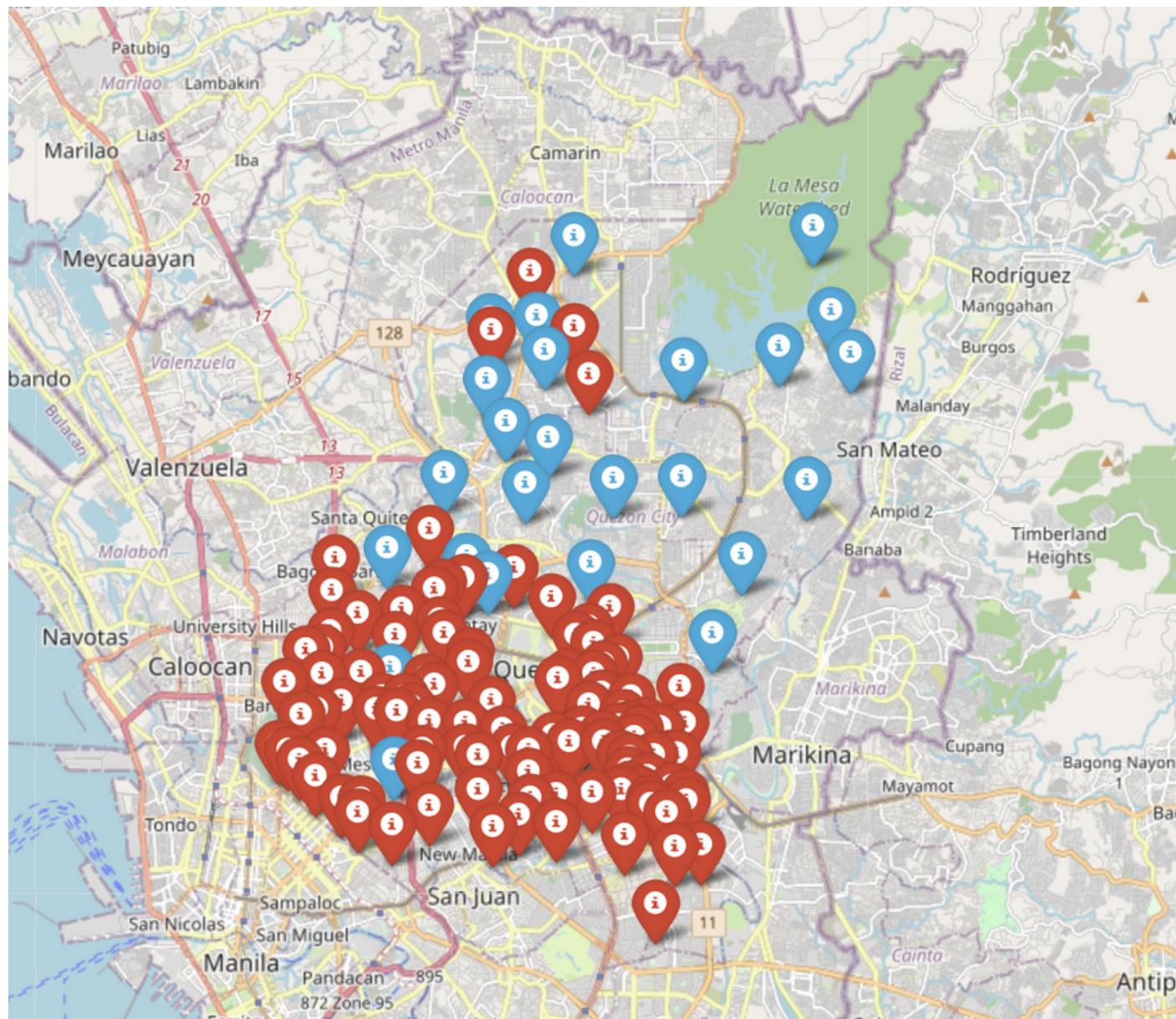
The datasets per geolocation are feature-engineered and fed through to a machine learning model



Hotspot Detection

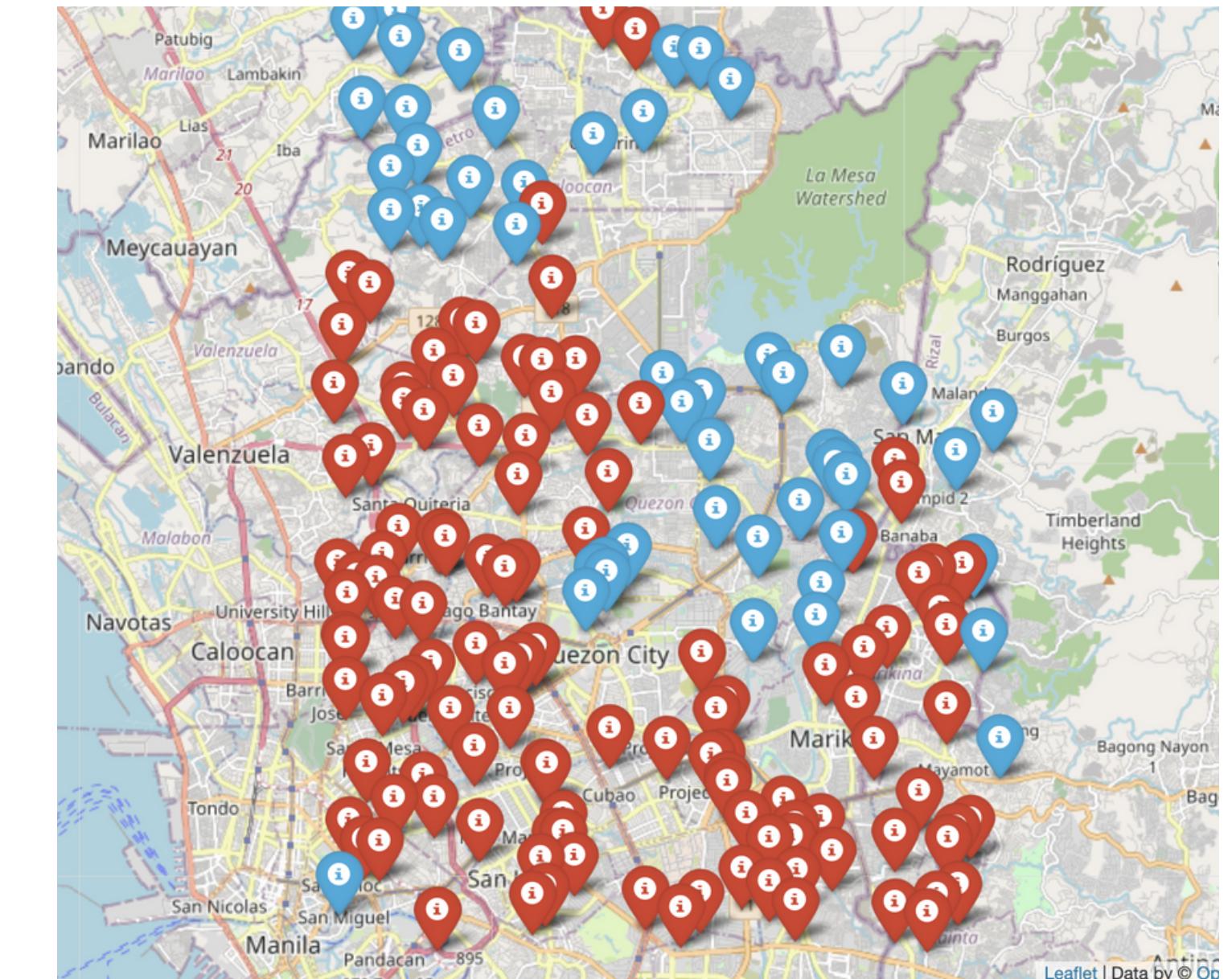
High Risk (>20 cases)

Low Risk (at least 1 case)



Ground Truth Hotspots

(From Philhealth (2016-2018 dengue cases
in Quezon City)



Modeled Hotspots

(Using remote sensing data,
vegetation index, water index, built-up index, air quality
index, surface temperature)

Web Application

QUEZON CITY, PHILIPPINES

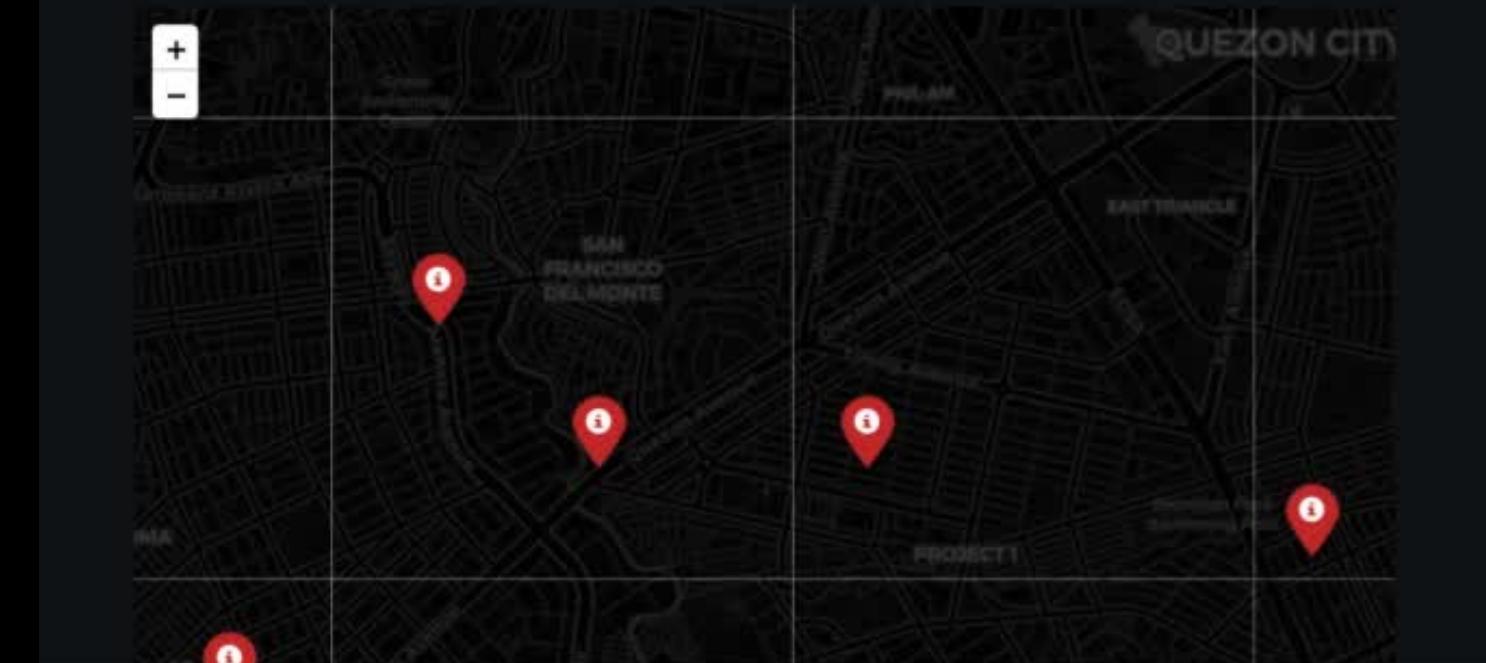
From just bounding box coordinates
of an area of interest, we're able to
identify potential at-risk locations for
dengue hotspots

AEDES: Predictive Geospatial Hostpot Detection

This web application demonstrates the use of satellite, weather and OpenStreetMap data to identify potential hotspots for vector-borne diseases. This web application only needs geojson input of an area of interest and then it automatically collects and models the data needed for hotspot detection at a longlat level.

Input geojson of area of interest here

```
[[[120.98976275,14.58936896],[121.13383232,14.58936896],[121.13383232,14.77641364],  
[120.98976275,14.77641364],[120.98976275,14.58936896]]]
```



Web Application

LAGUNA, PHILIPPINES

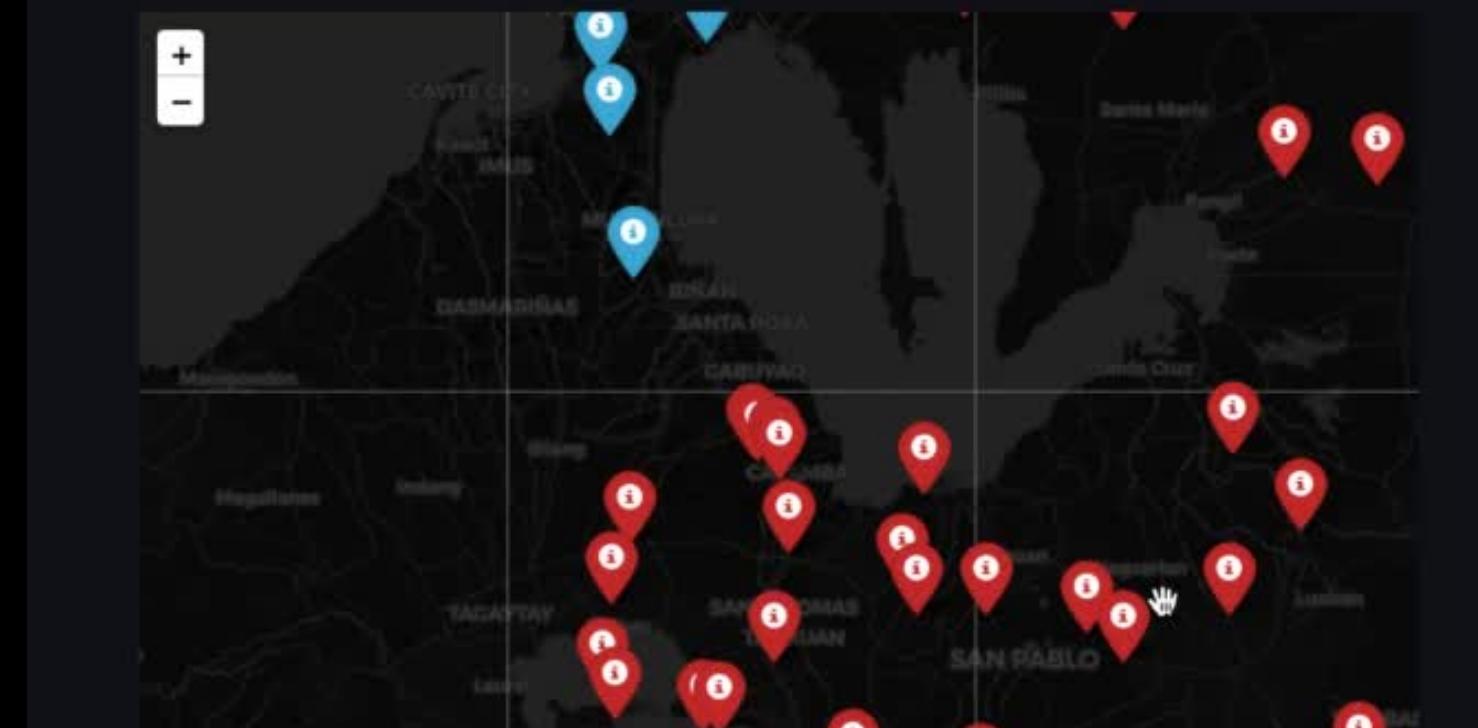
From just bounding box coordinates of an area of interest, we're able to identify potential at-risk locations for dengue hotspots

AEDES: Predictive Geospatial Hostpot Detection

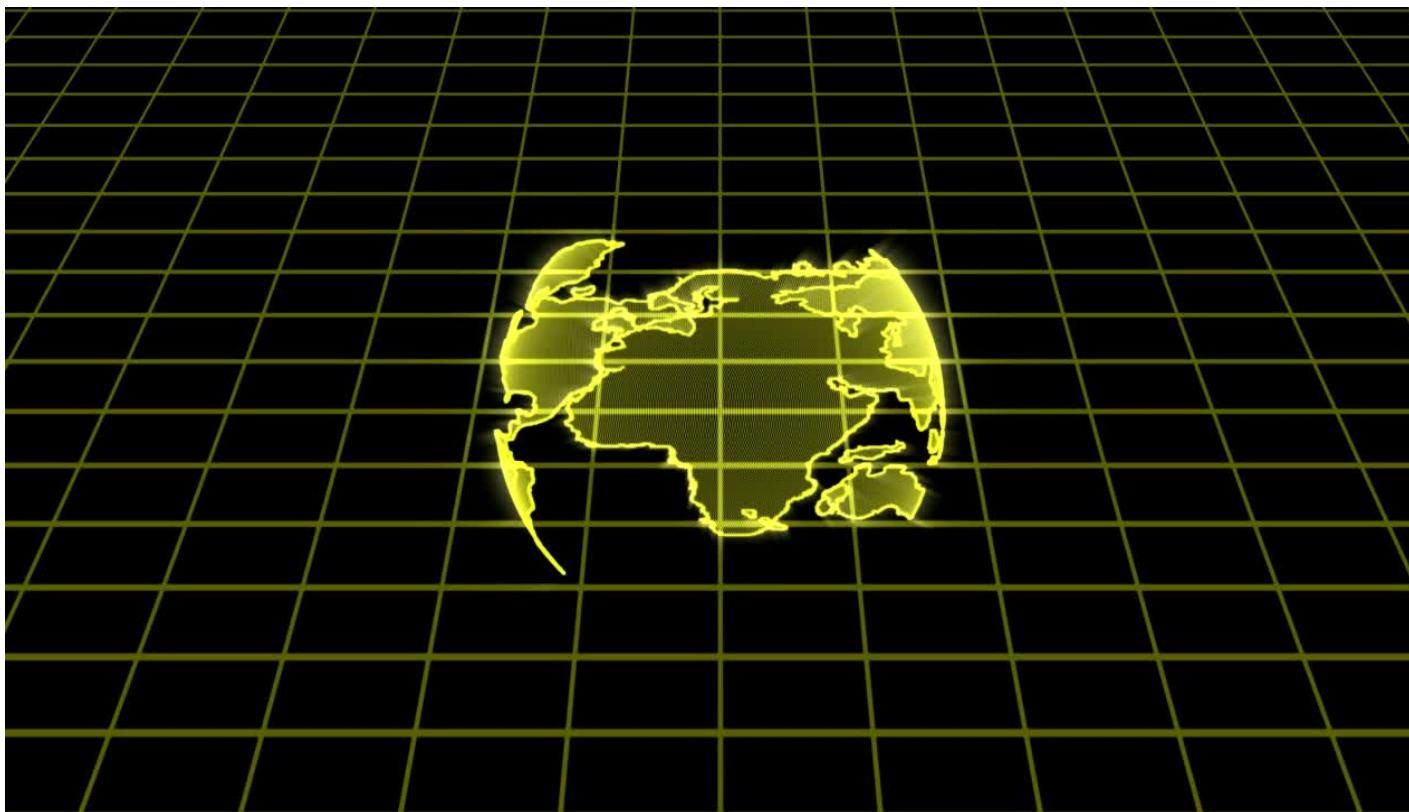
This web application demonstrates the use of satellite, weather and OpenStreetMap data to identify potential hotspots for vector-borne diseases. This web application only needs geojson input of an area of interest and then it automatically collects and models the data needed for hotspot detection at a longlat level.

Input geojson of area of interest here

```
[[[121.00197333,13.96948967],[121.59683223,13.96948967],[121.59683223,14.56562999],  
[121.00197333,14.56562999],[121.00197333,13.96948967]]]
```



ML Inference Input



Spatial

longitude

latitude

ISO geo tag (for search)

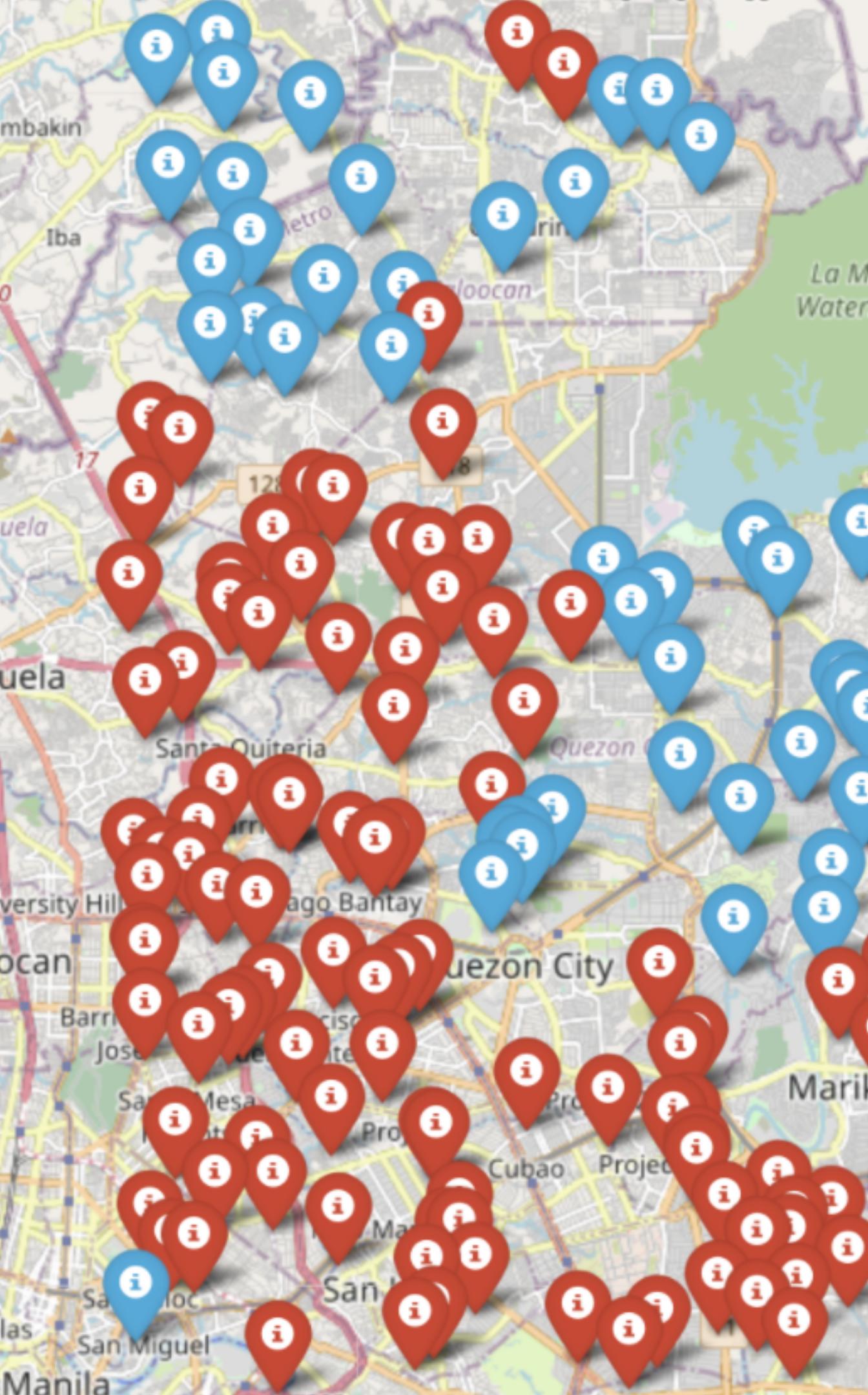


Temporal

Year

Month

Day



DATA CONSIDERATIONS

CURRENT DATA AUTOMATION

SATELLITE DATA TAKES TIME TO QUERY
~9 hours for 4.8k data points

OSM DATA IS MEMORY INTENSIVE
Contraction hierarchies crashes a 16 GB RAM
machine on 4.8k data points

PYTRENDS LOCATION ISN'T GRANULAR
ISO geo tags are used to limit search to a
location

DATA BACKLOG

TWITTER DATA

As discussed with BOA, sentiment analysis and NLP are needed to make sense of tweets vs simple engagements/likes/count

NIGHT LIGHTS

Implementation is not straightforward and will be added during future enhancements

FB DATA

BOA has streaming FB data for mobility and will be added when granted access during future enhancements