

Air Pollution

*in*

LOS ANGELES

Team: Lorelei

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# MOST POLLUTED CITIES IN THE UNITED STATES



## Ozone-polluted cities

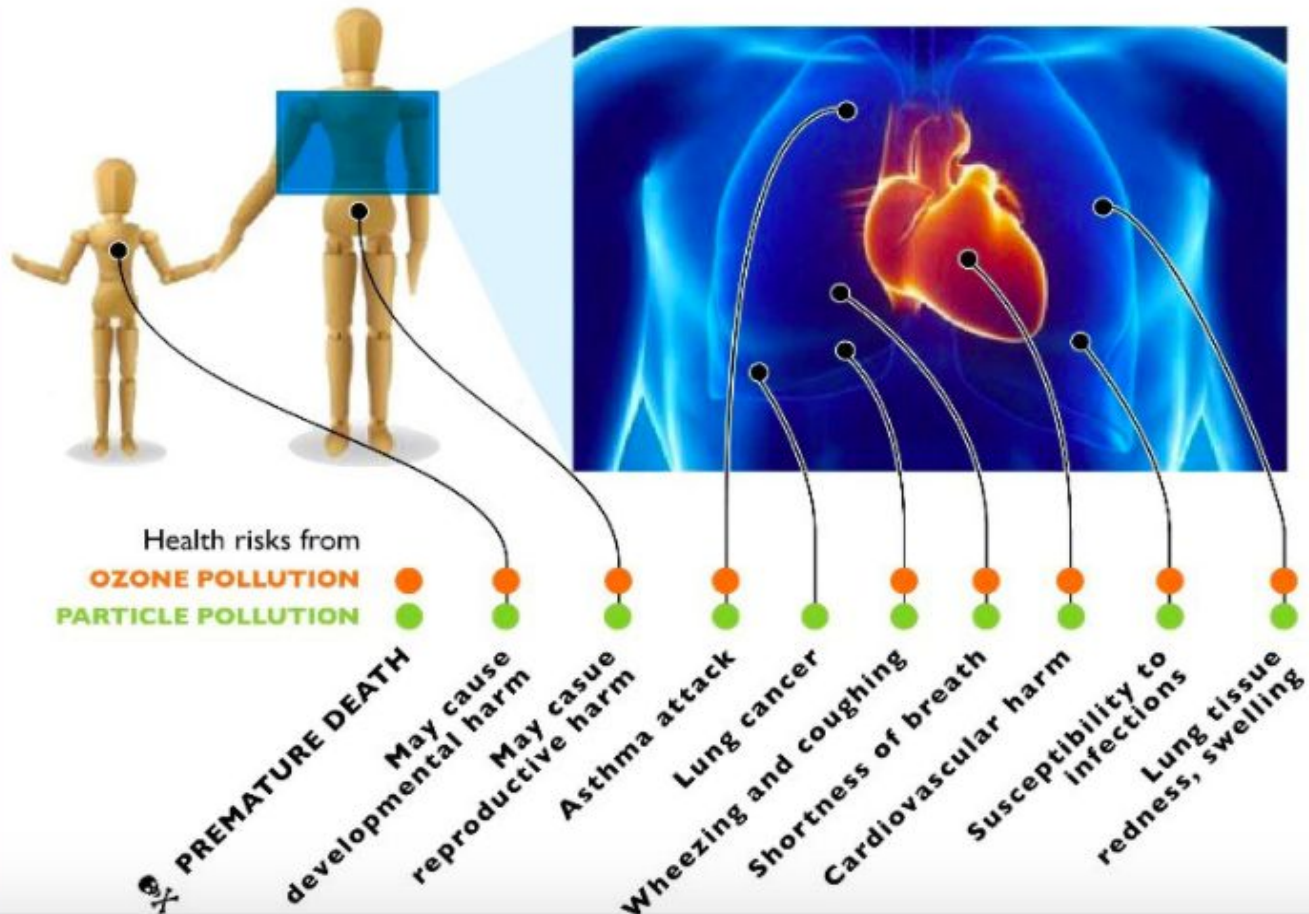
1	Los Angeles, Cali.
2	Visalia, Cali.
3	Bakersfield, Cali.
4	Fresno, Cali.
5	Sacramento, Cali.
6	Houston, Texas
7	Dallas, Texas
8	Modesto, Cali.
9	Las Vegas, Nev.
10	Phoenix, Ariz.

## Year-round particle-polluted cities

1	Fresno, Cali.
2	Bakersfield, Cali.
3	Visalia, Cali.
4	Modesto, Cali.
5	Los Angeles, Cali.
6	El Centro, Cali.
7	San Jose, Cali.
8	Cincinnati, Ohio
8	Wilmington, Ind.
8	Maysville, Ky.
9	Pittsburgh, Penn.
9	New Castle, Ohio
9	Weirton, W. Va.
10	Cleveland, Ohio

*Statistics courtesy of a new study  
by the American Lung Association*

**Air pollution remains a major danger to the health of children and adults.**







AirNow Home >> California >> **Central LA CO**

#### Data and Forecasts courtesy of:

Antelope Valley Air Quality Management District, Bay Area Air Quality Management District, California Air Resources Board, Imperial County Air Pollution Control District, Mendocino County Air Quality Management District, Mojave Desert Air Quality Management District, Monterey Bay Unified Air Pollution Control District, Placer County Air Pollution Control District, Sacramento Metropolitan Air Quality Management District, San Diego County Air Pollution Control District, San Joaquin Valley Air Pollution Control District, San Luis Obispo County Air Pollution Control District, Santa Barbara County Air Pollution Control District, South Coast Air Quality Management District, Ventura County Air Pollution Control District, Yolo-Solano Air Quality Management District

Current AQI

Forecast

AQI Loop

More Maps



 Tribal Boundaries

The tribal boundaries shown here are provided by the Bureau of Indian Affairs and are intended to be used as a general spatial reference only. They are not a formal determination of tribal boundaries by the EPA.

#### Current Conditions

Data Not Available

#### Air Quality Forecast

Today

Tomorrow

Air Quality Index (AQI)

74

Moderate

**Health Message:** Unusually sensitive people should consider reducing prolonged or heavy exertion.

Data Not Available

#### AQI - Pollutant Details

Carbon Monoxide

15

Good

Nitrogen Dioxide

20

Good

Ozone

37

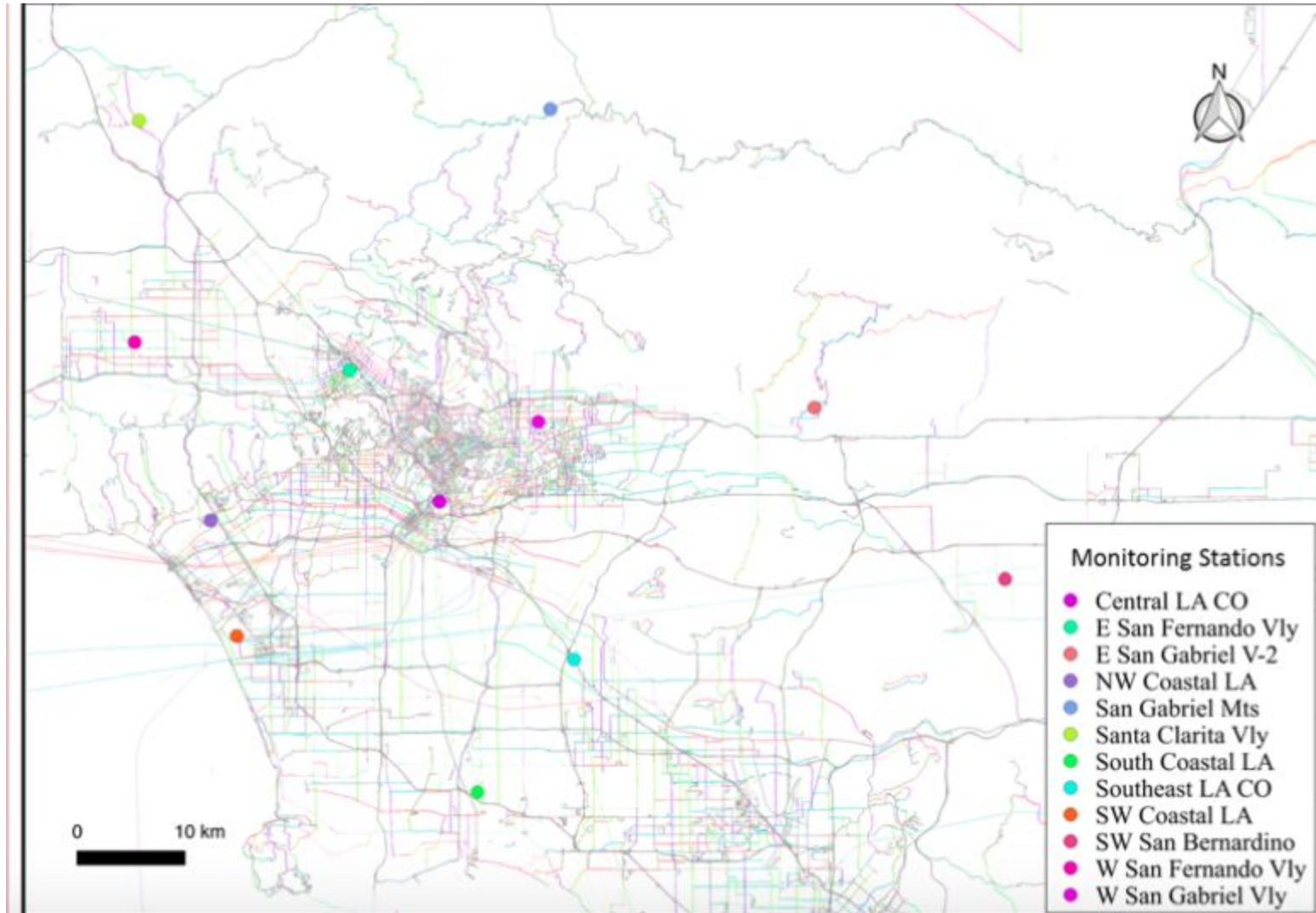
Good

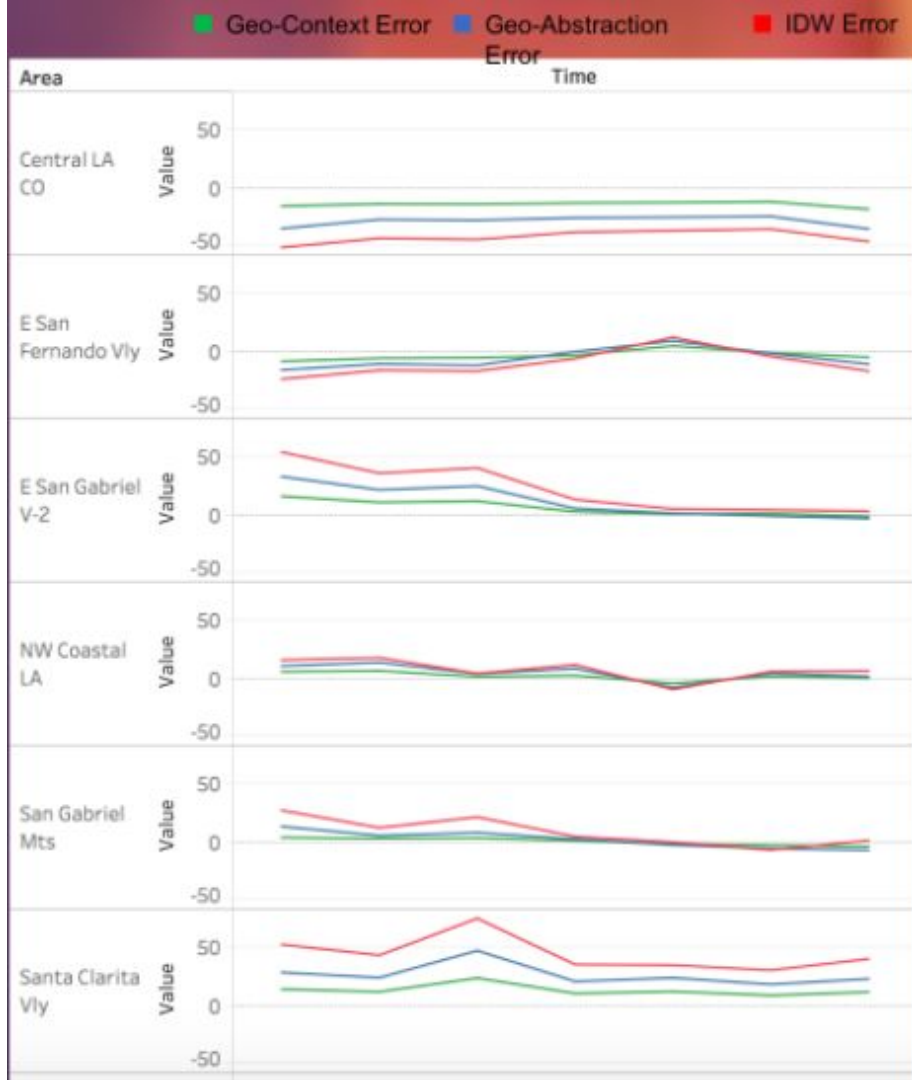
Particles (PM<sub>2.5</sub>)

34

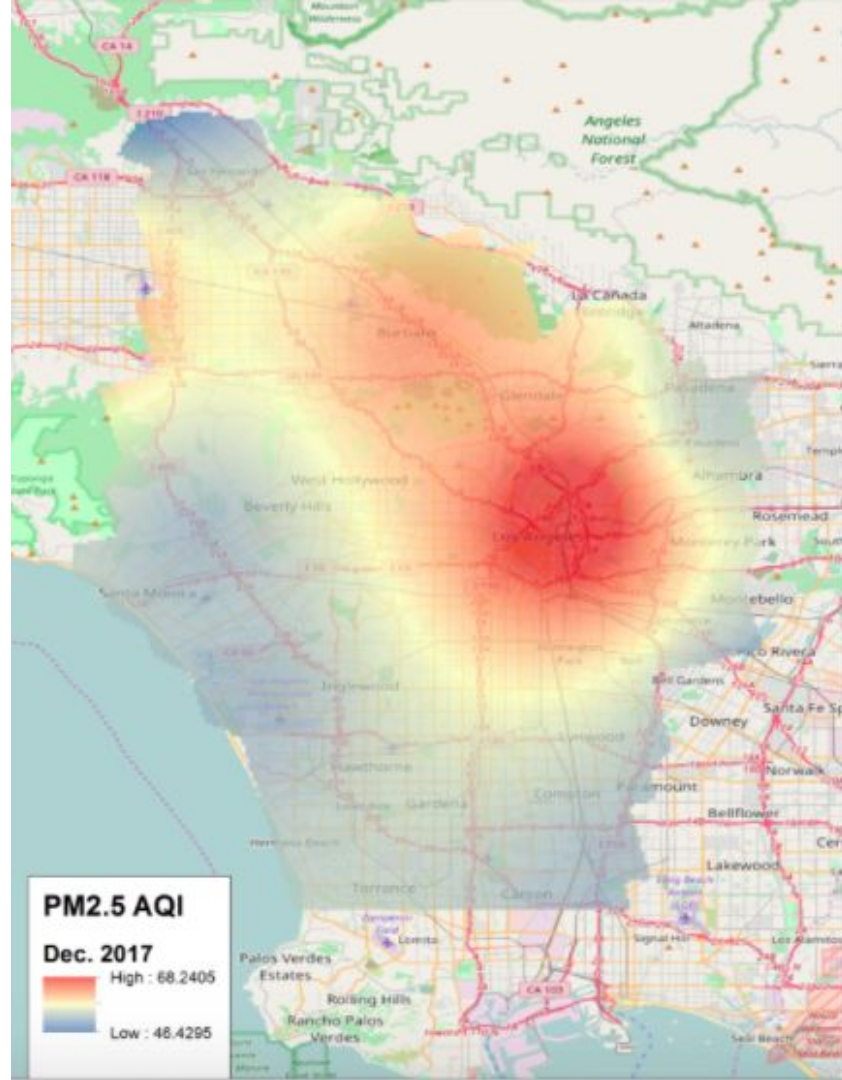
Good

Data Not Available

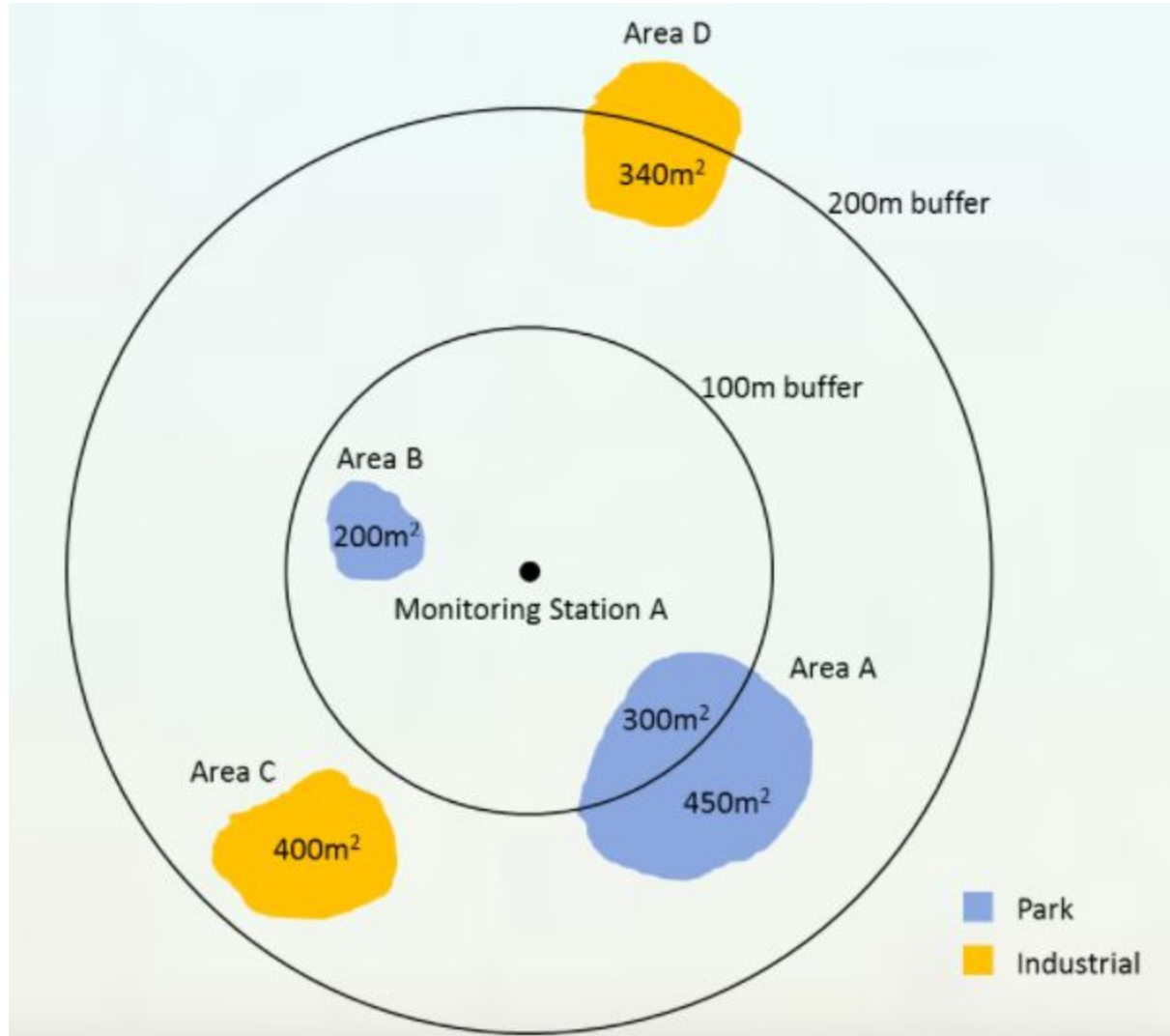


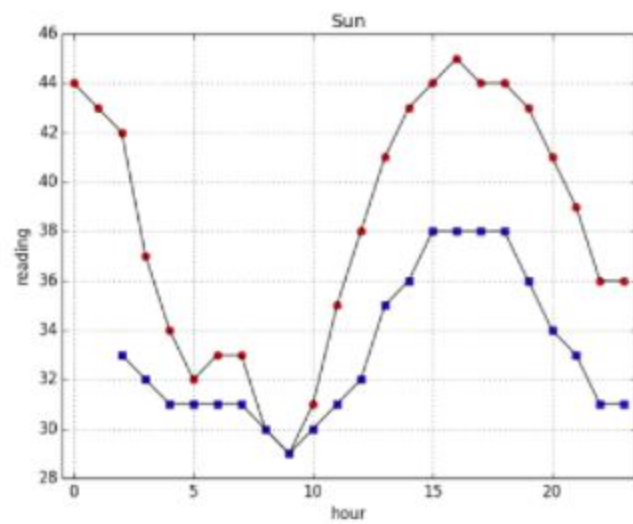
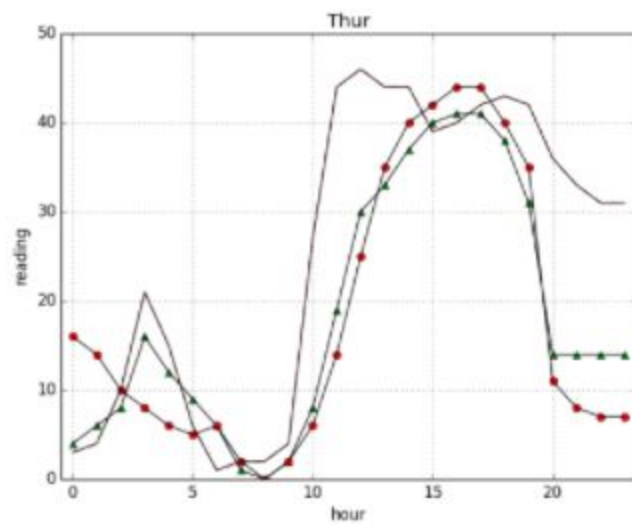












# Current AQI (Combined PM and O<sub>3</sub>)

Tuesday, October 10, 2017 11:00 PM PDT







# PROJECT TIMELINE

Data Cleaning



Exploratory Analysis



Prototyping



Deployment & Validation



Design Retouch



User Testing



Documentation & Presentation



Week 1

Week 2

Week 3

Week 4

**Fan Pan**  
data modeling &  
visualization programming

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aesthetics design &  
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## Mining Public Datasets for Modeling Intra-City PM<sub>2.5</sub> Concentrations at a Fine Spatial Resolution

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### ABSTRACT

Air quality models are important for studying the impact of air pollutant on health conditions at a fine spatiotemporal scale. Existing work typically relies on area-specific, region-selected pollution measures, dispersion and transport factors (e.g., meteorology) for building the model for each combination of study area, pollutant types, and spatiotemporal scales. In this paper, we present a data mining approach that utilizes publicly available OpenStreetMap (OSM) data to automatically generate an air quality model for the concentrations of fine particulate matter less than 2.5  $\mu\text{m}$  in aerodynamic diameter at various temporal scales. Our experiment shows that our OSM-based impact fine model could generate accurate PM<sub>2.5</sub> concentration estimates, which can be used to improve air quality models that traditionally rely on region-selected inputs. Our approach also quantifies the impact on air quality from a variety of geographic features (i.e., how various types of geographic features such as parking lots and commercial buildings affect air quality and how what distances representing inside, stationary and area sources and anthropogenic air pollution sources. This approach is particularly important for enabling the construction of context-specific spatiotemporal models of air pollution, allowing investigation of the impact of air pollution exposure on sensitive populations such as children with asthma at scale.

### CCS CONCEPTS

Information systems  $\rightarrow$  Spatial-temporal systems

### KEYWORDS

PM<sub>2.5</sub> Concentration, Air Quality Model, Geospatial Data Mining

### ACM Reference Format

Yipen Lin, Fan Pan, Yao-Yi Chung, Chia-Hsin Stripling, José Luis Arribas, Josée P. Eckel, and Rima Haber. 2017. Mining Public Datasets for Modeling Intra-City PM<sub>2.5</sub> Concentrations at a Fine Spatial Resolution. In *Proceedings of ACM SIGSPATIAL*. Atlanta, GA, USA, November 2017, 1002–1010. <https://doi.org/10.1145/3141111>

### 1 INTRODUCTION

Fine particulate matter (PM<sub>2.5</sub>) consists of particles less than 2.5  $\mu\text{m}$  in aerodynamic diameter that were inhaled can penetrate the respiratory region of the lungs and contribute to respiratory and cardiovascular diseases. Typical primary sources of contributions to fine particulate matter include stationary and moving vehicle exhausts, burning sources (e.g., wood-burning stoves and wildfires), refineries, and power plants. Secondary PM<sub>2.5</sub> is also formed as a result of photochemical reactions in the atmosphere in the presence of precursor gases and other catalysts. Epidemiological studies have shown associations between exposure to PM<sub>2.5</sub> and various health conditions, including lung and respiratory diseases [21], such as asthma [5, 11]. In the 1995 landmark air pollution “Harvard Six Cities Study” [14] and other recent studies [16], researchers reported associations between the levels of exposure to fine particulate concentrations and the risk of mortality and morbidity for cities all over the world. As a result of this scientific evidence and several other similar studies, many countries have set national ambient concentration standards for PM<sub>2.5</sub>. In the United States, the Environmental Protection Agency (EPA) set PM<sub>2.5</sub> standards and established the air quality index (AQI) to communicate relative health risk levels of current pollution levels compared to the standard, which is converted by PM<sub>2.5</sub> concentration. The US AQI ranges from 0 to 500 and consists of six categories: “Good”, “Moderate”, “Unhealthy for Sensitive Groups”, “Unhealthy”, “Very Unhealthy”, and “Hazardous”. From an outdoor pollution exposure index, one can calculate the corresponding AQI and its health risk category for each type of respiratory air pollution. For example, a 30  $\mu\text{g}/\text{m}^3$  PM<sub>2.5</sub> measurement corresponds to an AQI of 10 and is in the “Moderate” category, which means that “the quality is acceptable; however, for some pollutants there may be a moderate health concern for a very small number of people who are extremely sensitive to air pollution.” A 60  $\mu\text{g}/\text{m}^3$  PM<sub>2.5</sub> measurement corresponds to an AQI of 150 and is in the “Unhealthy” category.

Untitled

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data pipeline &  
visualization programming

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web development &  
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