Fostering Participation in Air Quality Studies Using Small Sensors

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Sources of Air Pollution



Mobile Sources (vehicles)

Stationary Sources (power plants, factories)



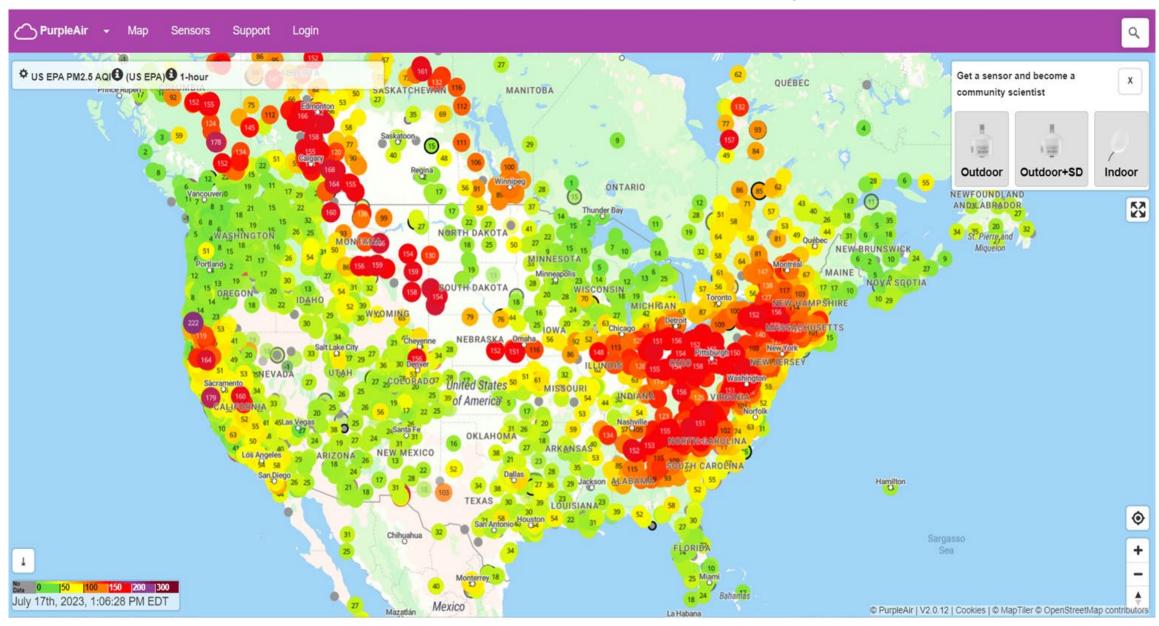
Area Sources (drycleaners, gas stations)

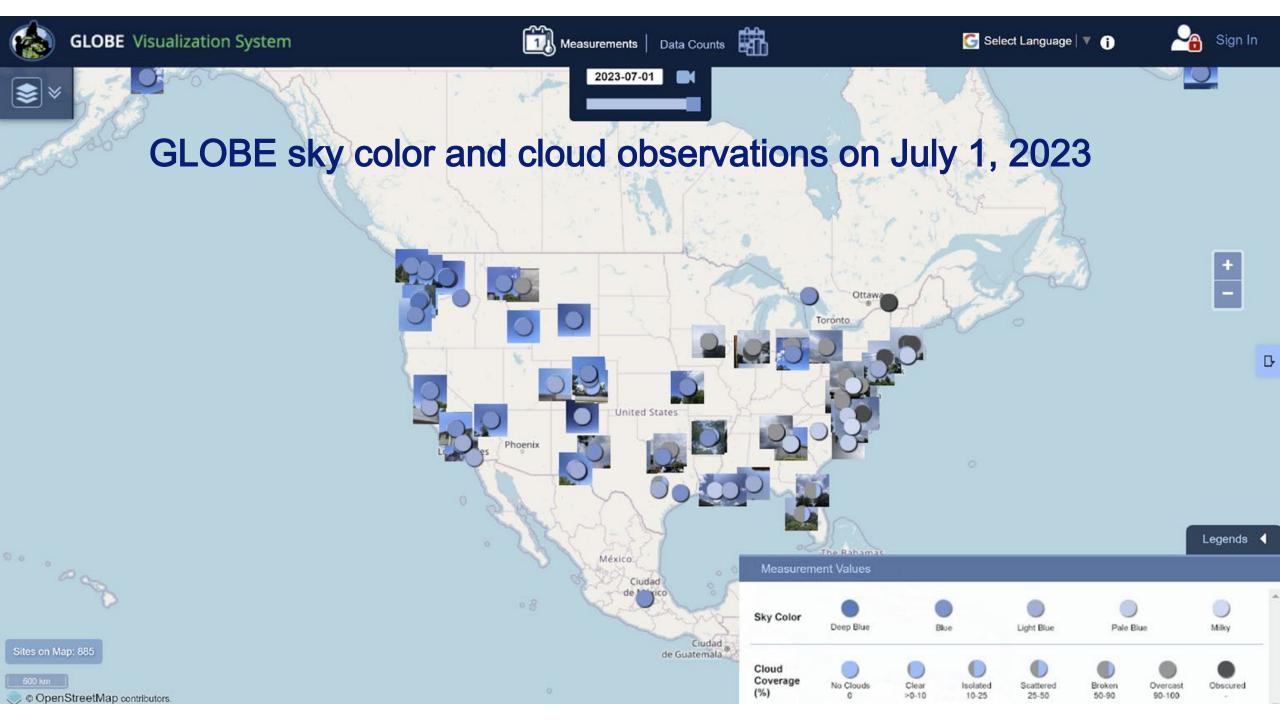


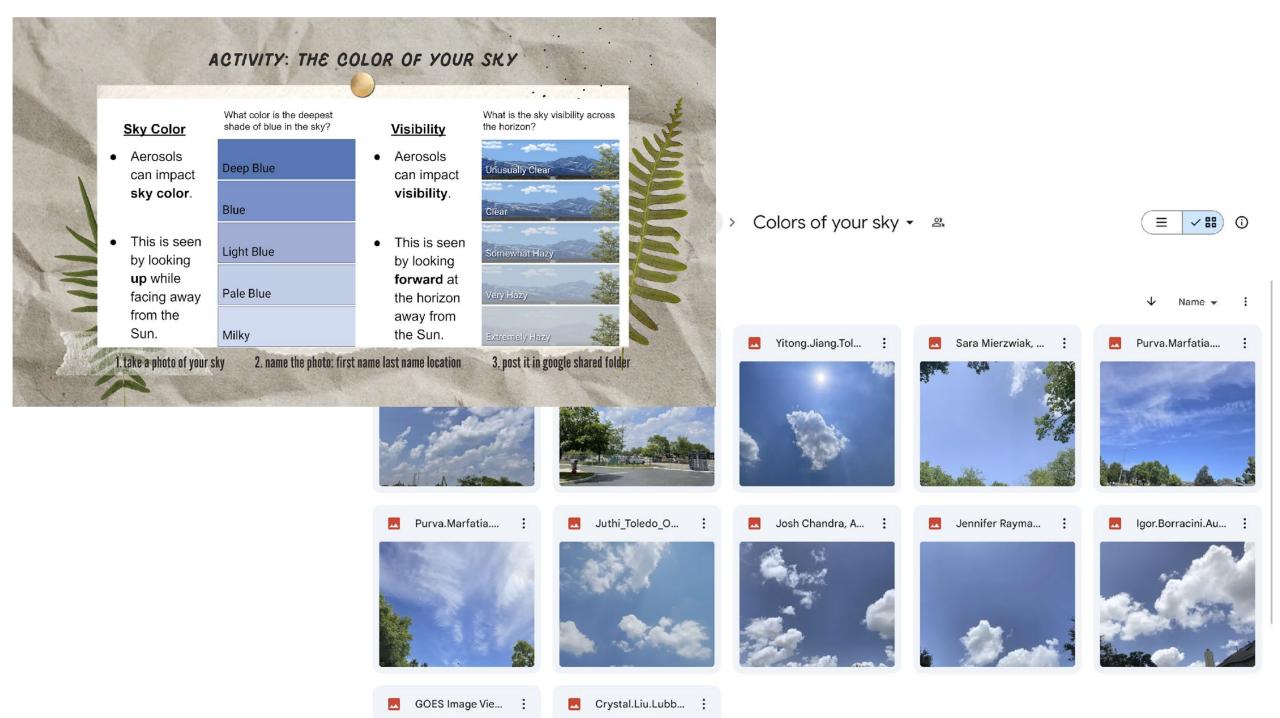


Natural Sources (forest fires, volcanoes)

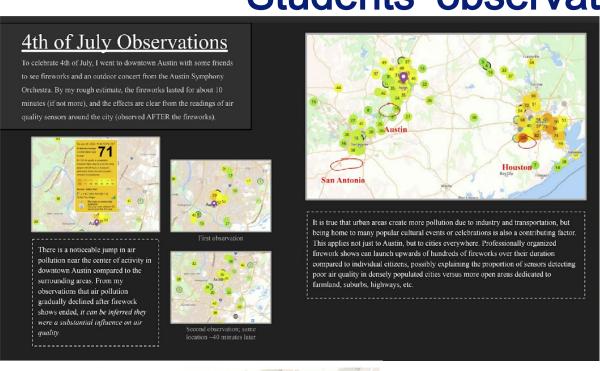
Concentration of PM 2.5 on July 1, 2023

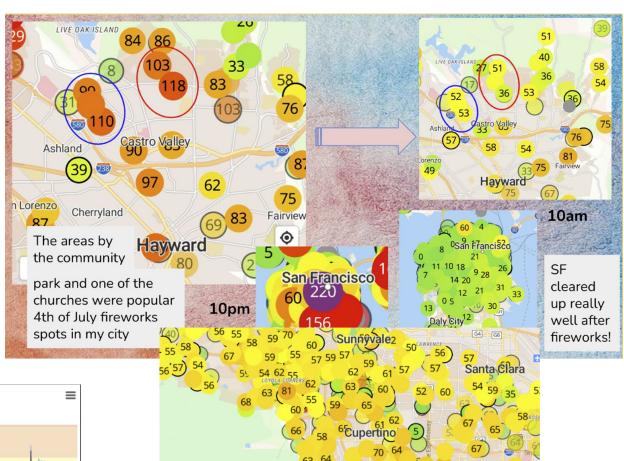






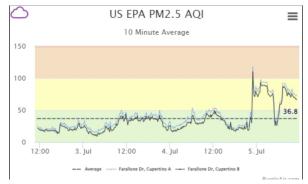
Students' observation on July 4, 2024







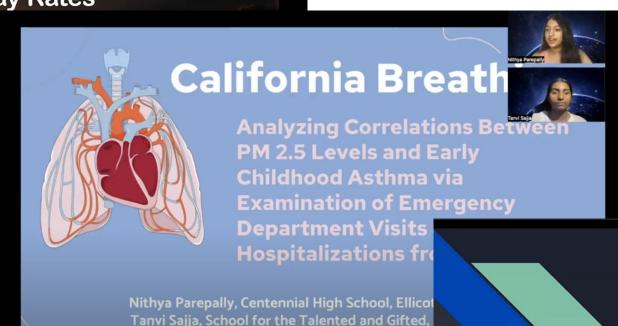






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NASA SEES Air Quality group research in 2023



Determining the Correlation between Economic Factors and Air Quality

ShawnEvan Gutierrez Advanced Technologies Academy

Lan Marias Nevada

A Study of the Las Vegas Valley

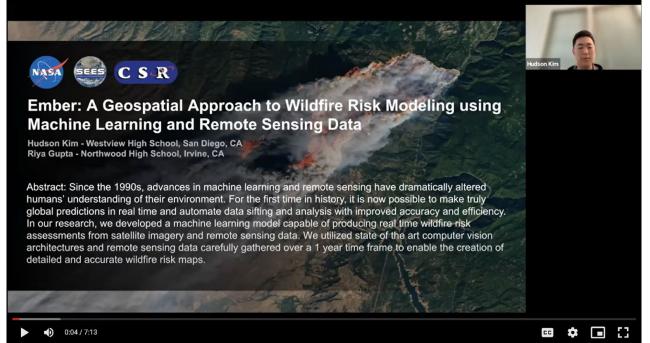












NASA SEES Air Quality group research in 2023 (cont.)

Continued research for publications

A Geospatial Approach to Wildfire Risk Modeling using Machine Learning and Remote Sensing Data

Riya Gupta, Northwood High School, Hudson Kim, Westview High School

Abstract—In recent years, the likelihood of wildfire occurrence has increased in many North American communities as changes in climate have led to longer, more deadly fire seasons. Many Americans, especially those living in Western states, have reported frequent drought and wildfire conditions, leading to an increased need for a modeling program to assess wildfire risk at a low computational cost. The research objective of this paper was to develop a machine learning model capable of producing accurate wildfire risk assessments using five geospatial datasets: Land Fire Mean Return, Annual Precipitation, Sentinel-2 Imagery, Land Cover, and Moisture Deficit & Surplus. To create the model, three separate machine learning architectures were implemented (U-Net, DeepLabV3, and the Pyramid Scene Parsing Network) and then applied to the study area of San Bernardino County, CA for the year of 2020. This study demonstrated a proof of concept for further inquiry into combining artificial intelligence and geospatial datasets to create useful insights.

■ Index Terms— Optical data, Geophysical data, Deep learning, Natural disasters and hazard

I. Introduction

SINCE California's first recorded wildfire in 1889, there has been a significant increase in the frequency of wildfires throughout the state. The increasing risk of wildfires has contributed to higher levels of inhalable particulate matter, placing an economic and health strain on local communities [1]. As wildfires continue to threaten California residents, accurate wildfire prediction models are crucial for managing fire response assets and addressing wildfire trends. Moreover,

accurate, and easily available to communities with frequent fire occurrence [3].

II. LITERATURE REVIEW

In recent years, the development of high-resolution satellite imagery from sources such as the Aqua and Terra satellites, has opened the possibility of developing an efficient and accurate wildfire risk modeling framework [4], [5]. Such frameworks support wildfire preparedness, one of the four components of wildfire management, through greater satellite data accessibility to support the development of a predictive model [6]. When taking data from a satellite source and applying it to a geospatial machine learning model, the accuracy of the results depends on the quality of the datasets used to train the model. Scott et al. presented an open-source Wildfire Hazard Potential map which provides further research into wildfire risk forecasting based on population density, building coverage, housing unit density, building exposure type, housing unit exposure, housing unit impact, and housing unit risk updated yearly [7]. With its results, this dataset provided promise in being used as a ground truth data layer to help train a model with similar input factors.

Several studies have been conducted using various physics-based parameters and environmental factors to model the growth and spread of fire during a wildfire event. In June 2022, a study conducted in London used inverse modeling based on Latent Assimilation techniques to estimate the parameters of a wildfire, finding that machine learning driven prediction models created a precise reconstruction of test datasets [8]. Additionally, a number of studies have applied

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Assessing the impact of wildfire smoke on PM2.5 levels: comparing outdoor and indoor air quality

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11 Student Authors

12 Please list all student author names here and indicate whether in middle or high school

13 Rebekah Repp – High School

15

16 SUMMARY

17 The 2023 wildfires in Canada have led to increases in PM2.5 levels throughout the U.S.

18 since early May 2023. The purpose of this study is to evaluate the impact of outdoor

19 wildfire smoke on indoor PM2.5 levels. Outdoor and indoor air quality was monitored

20 through changes in PM2.5 particles from PurpleAir sensors from July 1 to July 14. Data

21 was collected from five sensors in four different locations in Sylvania, Ohio; San

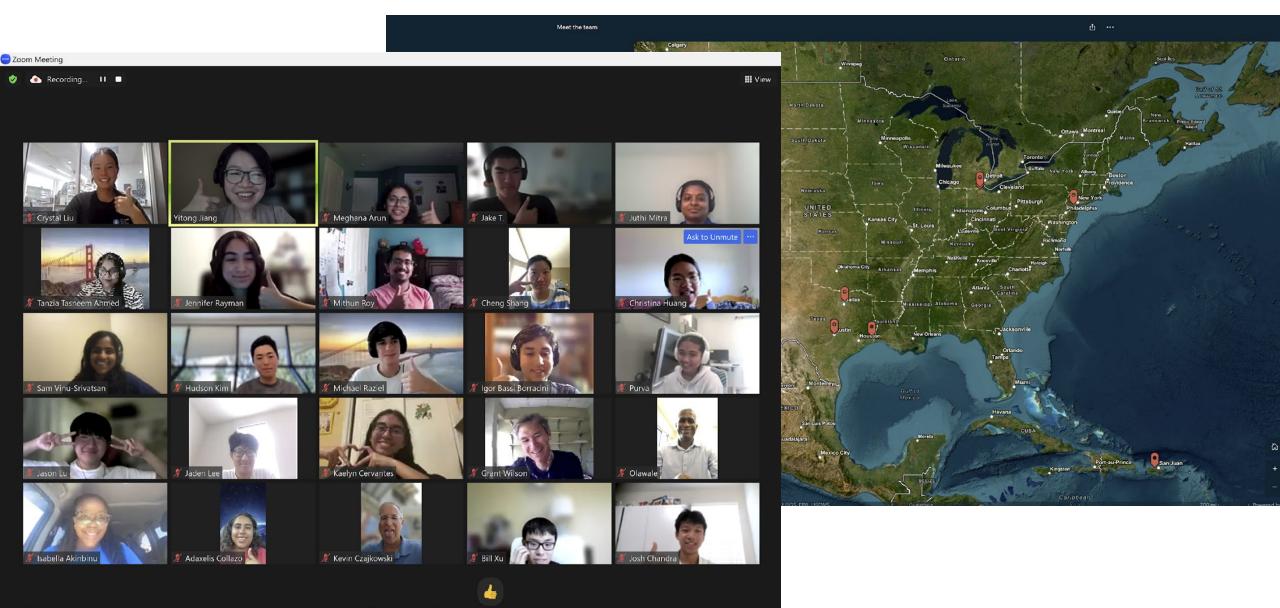
22 Bernardino, California; New York City, New York; and Chicago, Illinois; with two sets of

23 sensors located in Chicago, each with an indoor and outdoor sensor within 50 feet of

24 each other. The I/O ratios (the indoor PM2.5 amount divided by the outdoor PM2.5

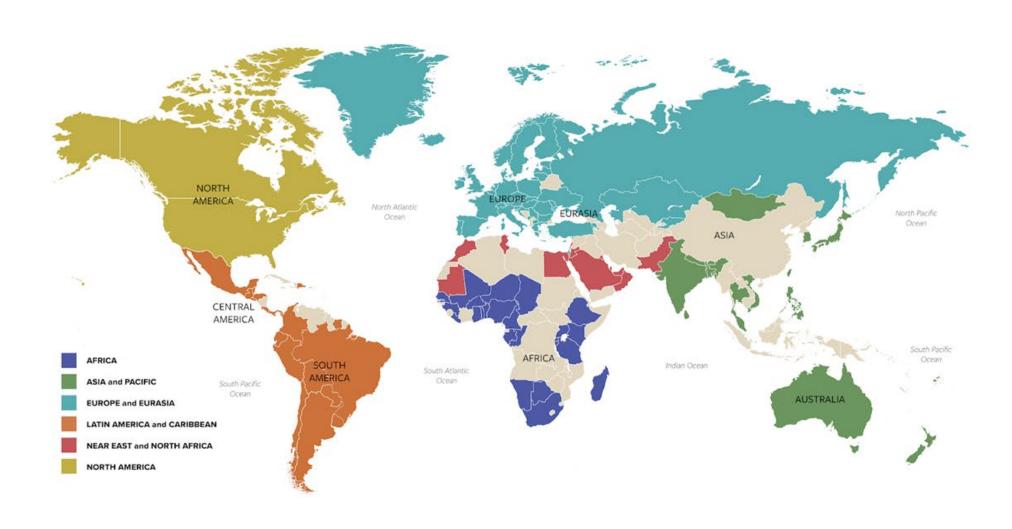
25 amount) of the mean indoor and outdoor PM2.5 levels and R-values of July 1-July 14

Air Quality Interns 2024





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