

USING MACHINE LEARNING AND  
GOOGLE EARTH  
ENGINE TO UNDERSTAND LAND USE  
AND LAND COVER  
CLASSIFICATIONS AND NO<sub>2</sub> LEVELS  
IN CALIFORNIA

Author(s): Benyamain Yacoob, Ethan Scheys, Eyiara Oladipo, Andre Price  
Advisor: Dr. Shadi Banitaan  
University of Detroit Mercy  
College of Engineering and Science (ECECS)



# Overview

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- Introduction
- Methodology
- Dataset(s)
- Results
- Implications

# Introduction

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- **Picture This:** Harnessing satellite imagery and AI to decode the air we breathe and revolutionize urban planning
- **Why It Matters:** Unveiling the hidden connections between how we use land and the quality of our environment
- **Join Us:** Discover the methods that could transform our cities into healthier, greener spaces

# Methodology

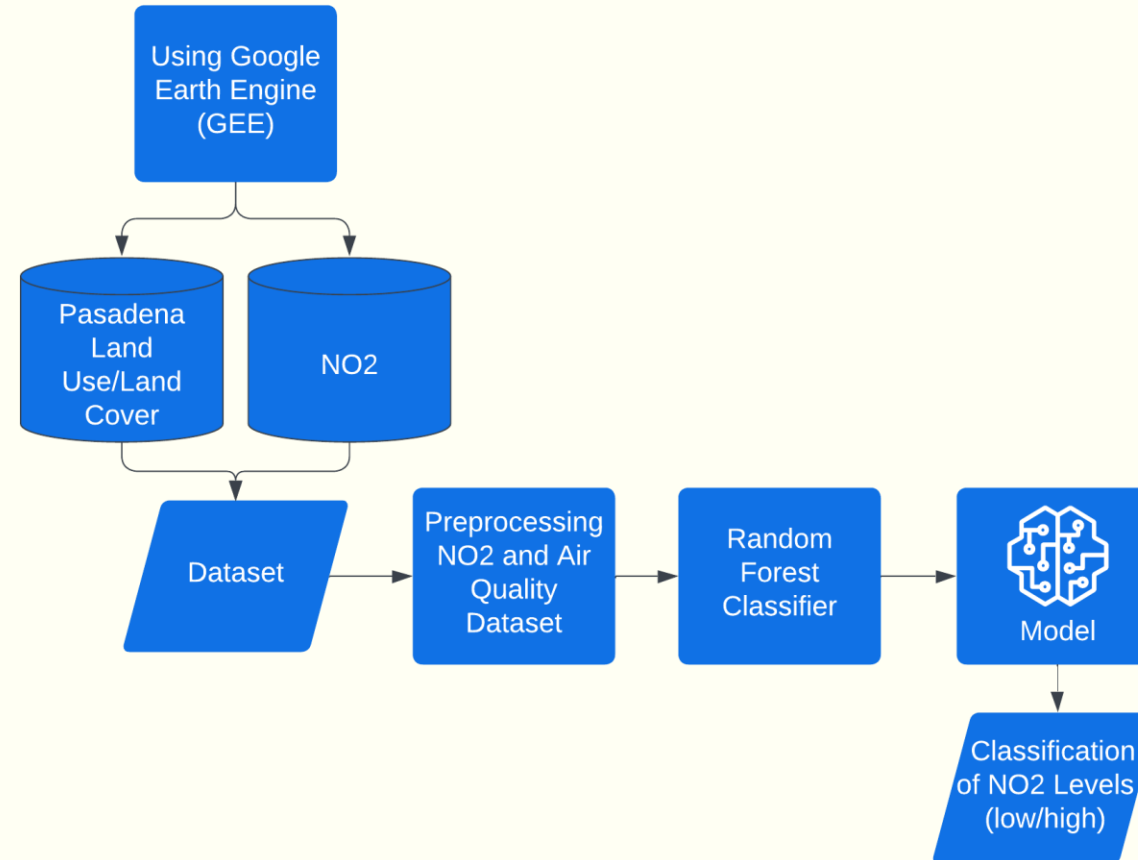
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- Focused on Pasadena, California as the study area
- Used Google Earth Engine (GEE) to prepare land cover and air quality datasets
- Sampling strategies and data distribution across attributes were described
- Random Forest (RF) and Decision Tree (DT) classifiers were used for analysis



# Methodology

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# Dataset(s)

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- Land Cover Datasets:

- Used Google Dynamic World Dataset - GOOGLE/DYNAMICWORLD/V1
- Data filtered monthly from January 1, 2021, to December 1, 2021 (199 images)
- Used "sampleRegions" function to sample pixel values and obtained 1,048,576 elements

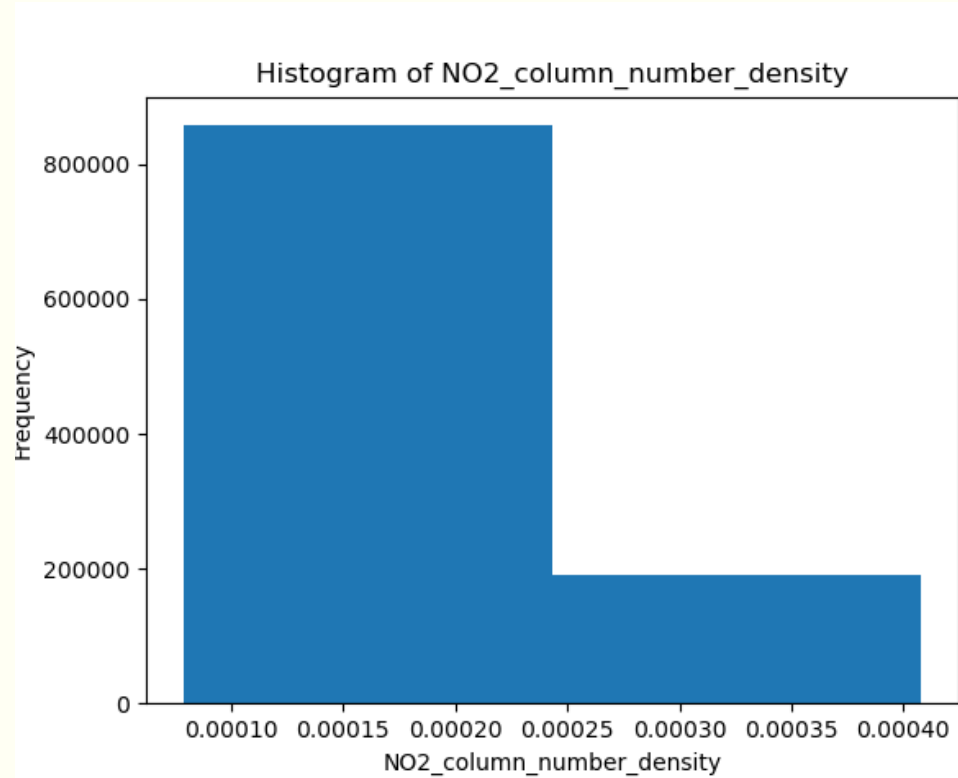
- Air Quality Datasets:

- Used Sentinel-5P NRTI NO<sub>2</sub> dataset for near real-time high-resolution imagery of nitrogen dioxide concentrations
- Used "NO<sub>2</sub> column number density" band (measured in mol/m<sup>2</sup>)
- Equal frequency discretization was used to create a binary classification task with "Low NO<sub>2</sub> Levels" and "High NO<sub>2</sub> Levels"

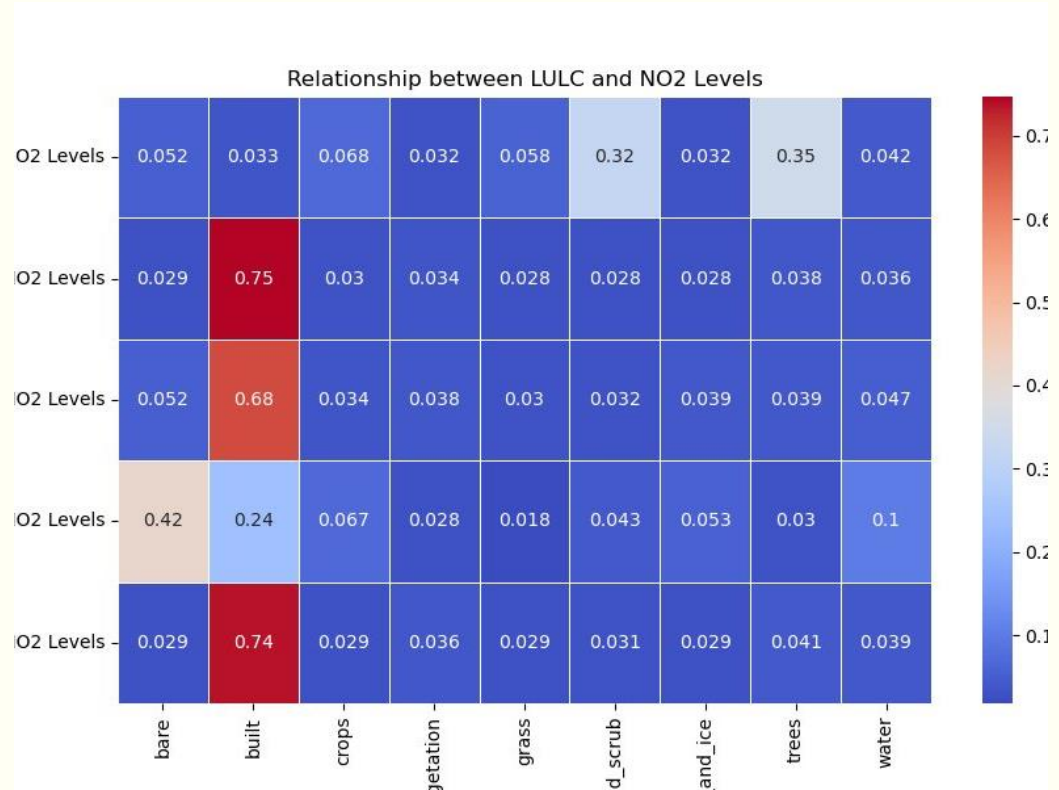
# Dataset(s)

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## Histogram



## Heatmap

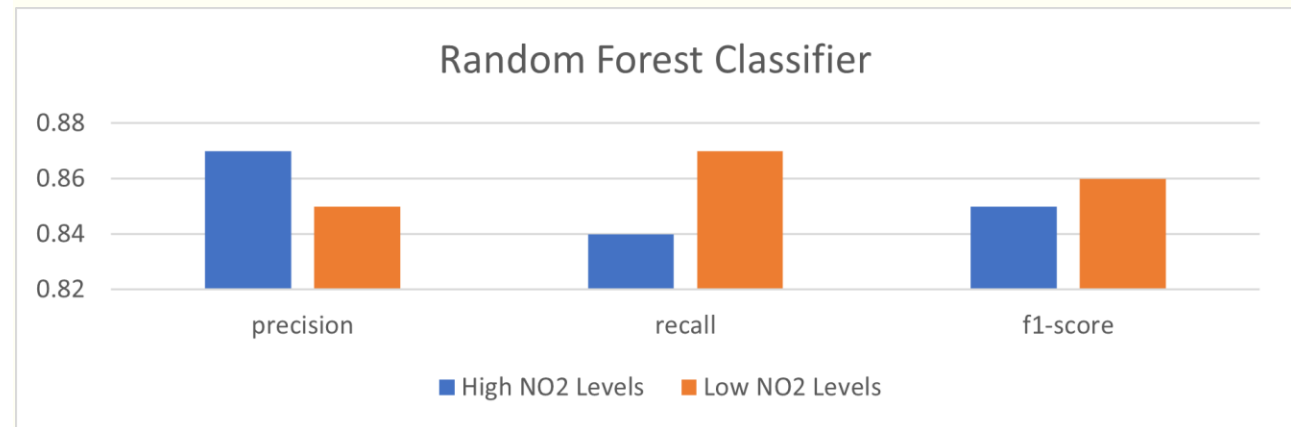
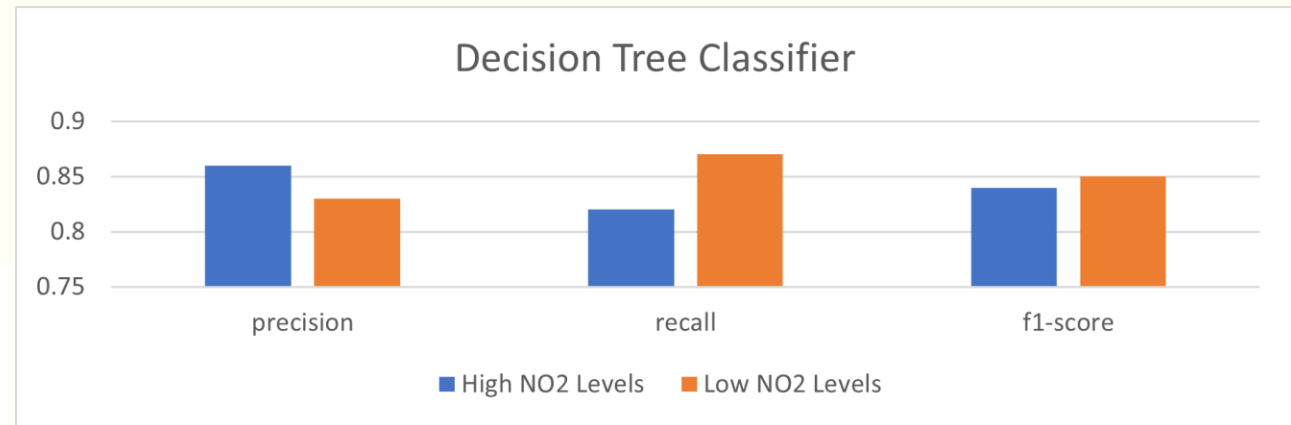


# Results

- Feature importances and classification report: Internal Python programming language library

TABLE I: Feature Importance of Attributes Retrieved from LULC.

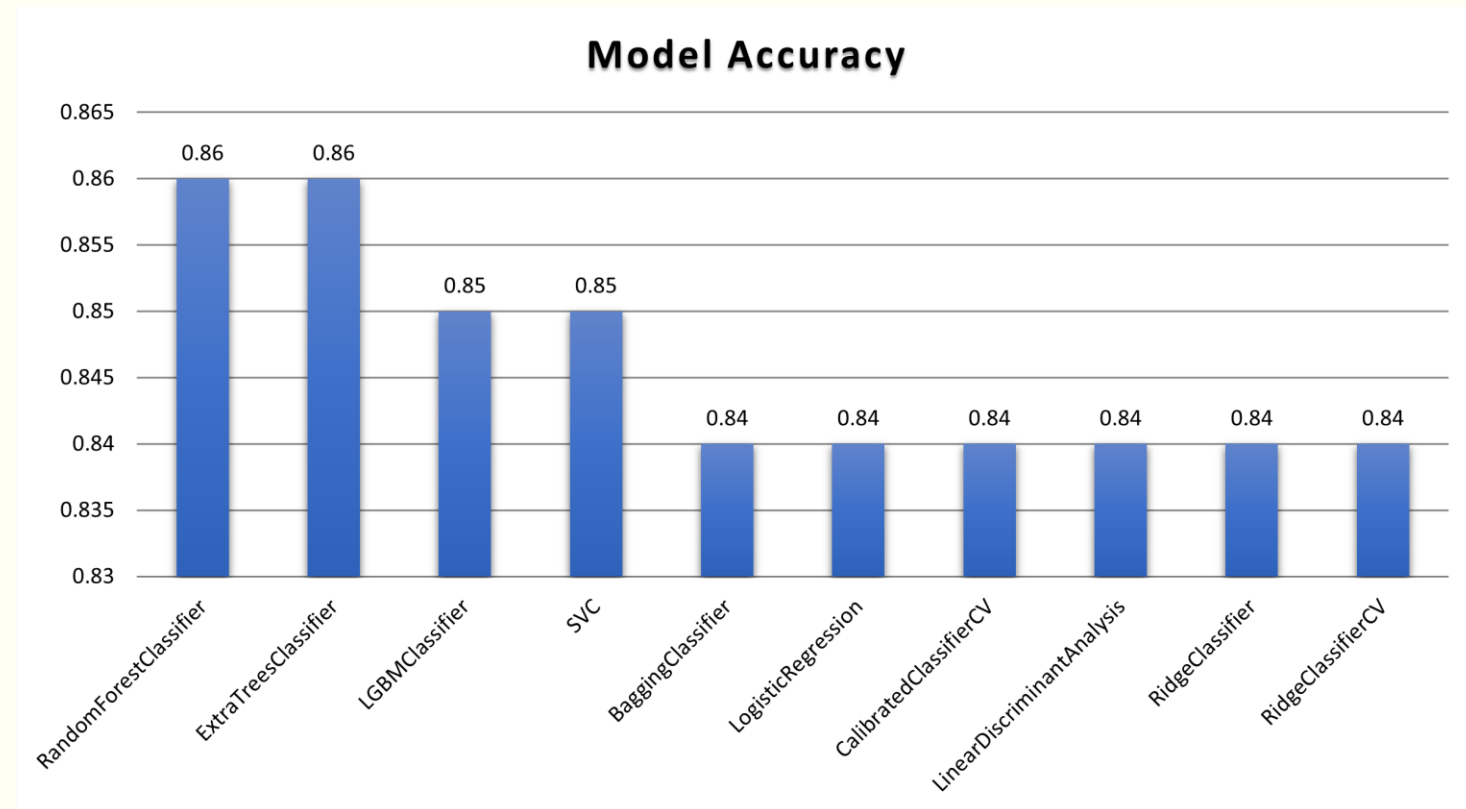
Attribute Type	Importance Score
Built	0.25
Shrub and Scrub	0.19
Trees	0.11
Grass	0.11
Water	0.07
Bare	0.07
Crops	0.07
Snow and Ice	0.06
Flooded Vegetation	0.06





# Results

- Decision Tree (DT) accuracy: 85%
- Random Forest (RF) accuracy: 86%
- 10-fold cross-validation was used, achieving a mean accuracy of 86%
- The strong association between LULC and air quality (specifically NO<sub>2</sub> presence) was concluded based on the high accuracy



# Implications

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- The presence of NO<sub>2</sub> diminishes quality of life and standard of living, affects production of crops
- Incorporate greenery spaces to counteract anthropogenic activities present in urban areas
- Different regions of interest (variety in natural landscapes, man-made structures)
- Environmental policy regulations

