Satellite Hotspot-Based Air Quality Forecasting



DEC 20, 2024



Agenda



- Introduction
- Problem Statement
- Procedures
- Finding
- Conclusion
- Recommendation
- Future Works

Introduction: The Silent Killer





Introduction: Sources of PM 2.5 Pollution

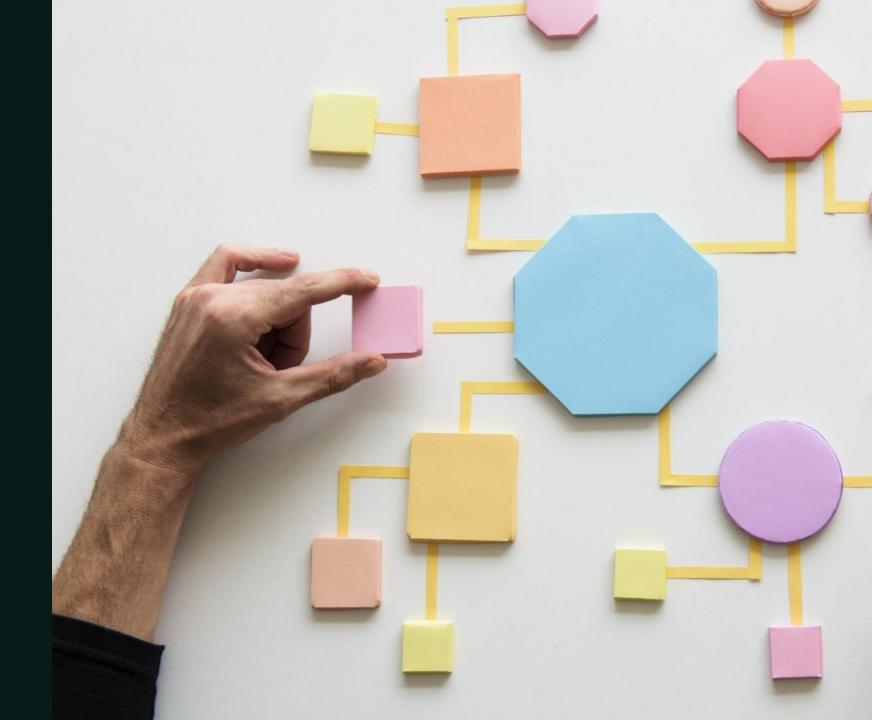
- Agricultural Burning: Open burning of agricultural waste, such as crop residue and rice straw
- · Forest Fires: Wildfires and forest fires, particularly during dry seasons
- Domestic Biomass Burning: Burning of wood, charcoal, and other biomass fuels for cooking and heating in rural areas
- Industrial Emissions: Industrial activities, especially those involving combustion processes
- Vehicle Emissions: Exhaust fumes from vehicles, particularly diesel-powered vehicles

Problem Statement

• Leveraging Satellite-Based and Weather Data for Daily Air Quality Forecasting in Thailand

Procedure





Data Sources

- Weather Data
 - Temperature (2m, 80m, 120m, 180m)
 - Relative Humidity (2m)
 - Dew Point (2m)
 - Pressure (MSL, Surface)
 - Wind Speed (10 m, 80 m, 120 m, 180 m)
 - Wind Direction (10m, 80 m, 120 m, 180 m)

Data Sources

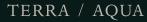
- Weather Data
 - Data Point: 6,076,656
 - Data Range: Sep 9, 2024 Dec 11, 2024
 - Coordinate: 13.35 97.05, 20.4 105.45
 - Source: DWD ICON via Open-Meteo
 - License Type: Commercial Subscription 29€/Month

- Air Quality Data
 - PM 2.5 (µg/m³) (Prediction Target)
 - UV Index

- Air Quality Data
 - Data Point: 6,106,752
 - Data Range: Data Range: Sep 9, 2024 Dec 11, 2024
 - Coordinate: 13.35 97.05, 20.4 105.45
 - Source: CAMS (Copernicus Atmosphere Monitoring Service) via Open-Meteo
 - License Type: Commercial Subscription 29€/Month

- Satellite Hotspot Data
 - Satellite
 - Aqua/Terra
 - SUOMI
 - NOAA 20, 21
 - Data Point:17,933
 - Data Range: Data Range: Sep 9, 2024 –
 Dec 11, 2024
 - Source: Fire Information for Resource
 Management System (FIRMS)







SUOMI

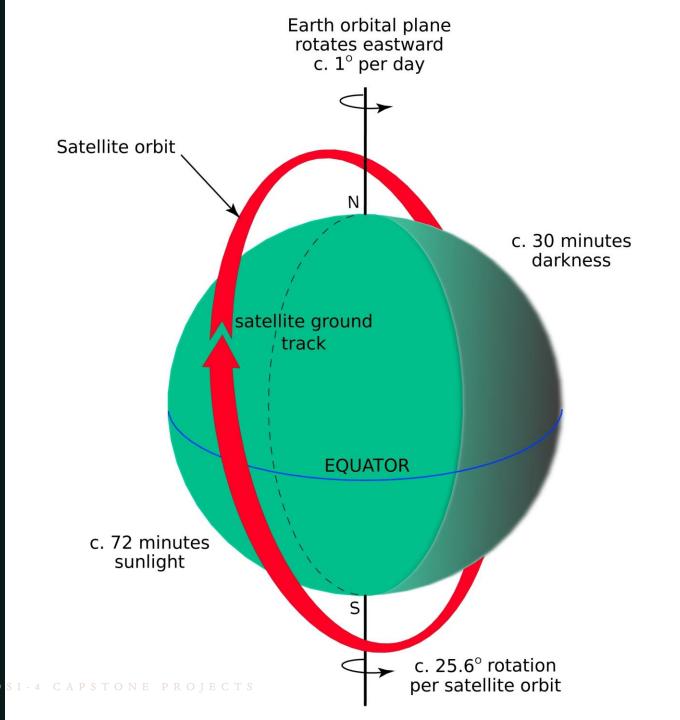


NOAA-20



NOAA-21

• Satellite Hotspot Data



Candidate Algorithms

- Classical Regression
 - Linear Regression
 - ElasticNet
 - XGBoost
 - Random Forest Regressor
- Deep Learning
 - Long Short-Term Memory (LSTM)

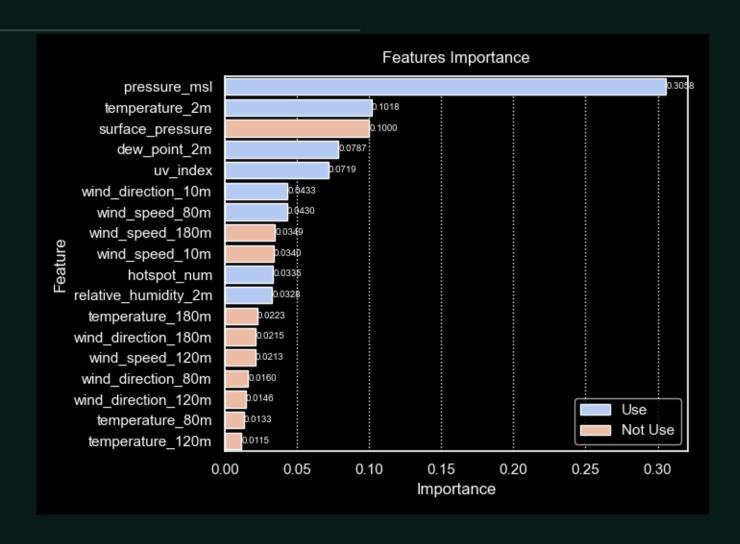
Findings





Features Importance

- 1. Pressure hPa (MSL)
- 2. Temperature C° (2m)
- 3. Dew Point C° (2m)
- 4. UV Index
- 5. Wind Direction Degree (10m)
- 6. Wind Speed km/h (80m)
- 7. Hotspot Number
- 8. Relative Humidity % (2m)

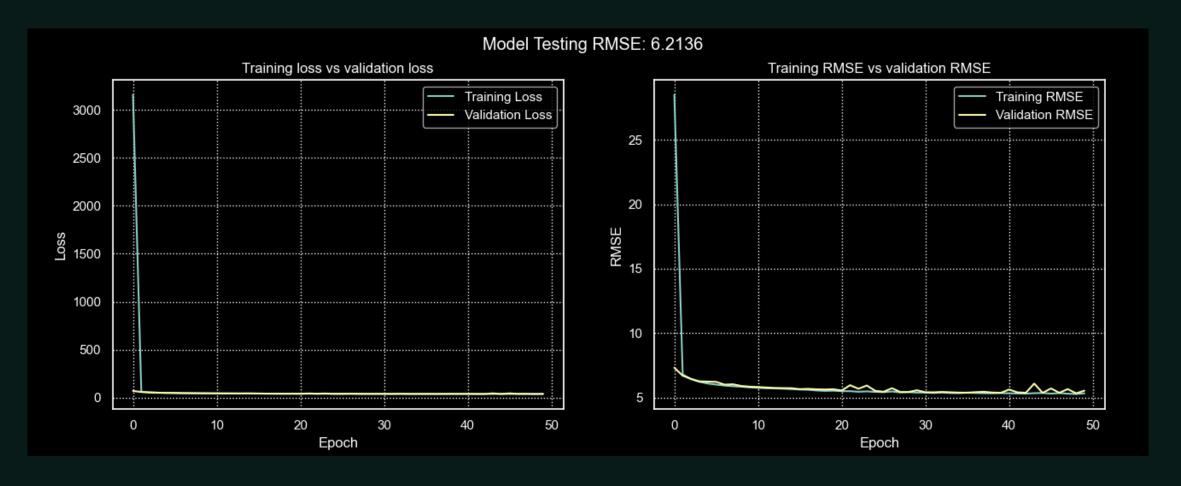


Models Comparison

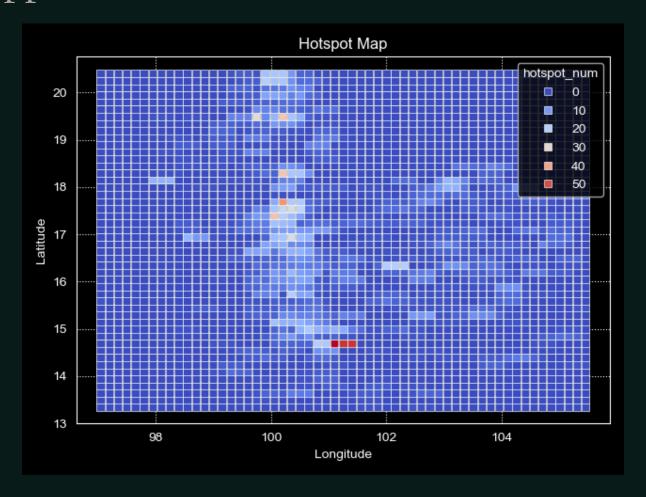
| Algorithm Name | Cross Validation Score | Train Score | Test Score | Test RMSE |
|---------------------------|---------------------------|-------------|------------|-----------|
| Random Forest | 0.83 | 0.98 | 0.82 | 2.71 |
| XGBoost | 0.62 | 0.64 | 0.59 | 4.15 |
| Linear Regression | 0.48 | 0.48 | 0.46 | 4.73 |
| ElasticNet | 0.46 | 0.46 | 0.44 | 4.82 |
| Long Short-Term Memory | | | | 6.10 |

• Select RandomForestRegressor since it has highest cross validation R² score

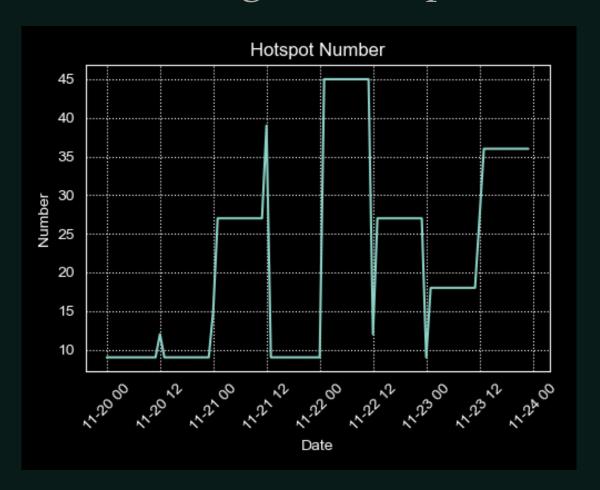
Long Short-Term Memory Model Graph

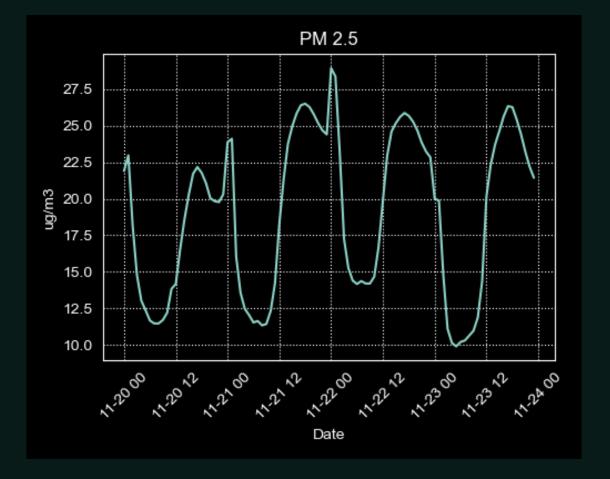


Hotspot of Upper Part of Thailand

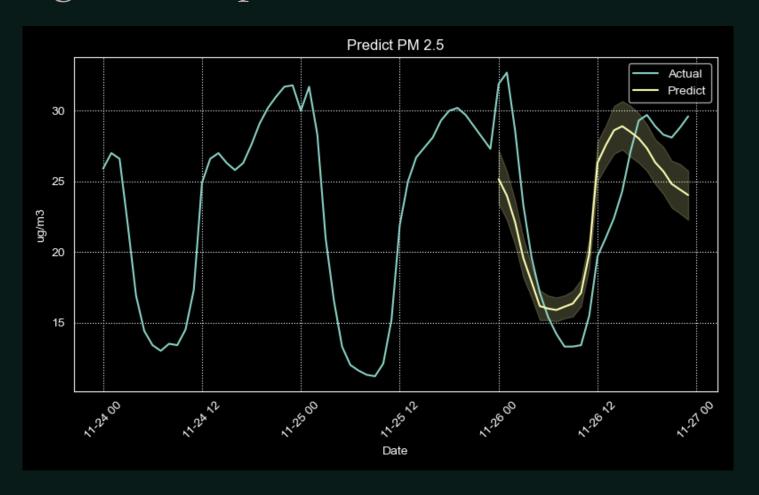


Case: Chiang Mai Hotspot Number

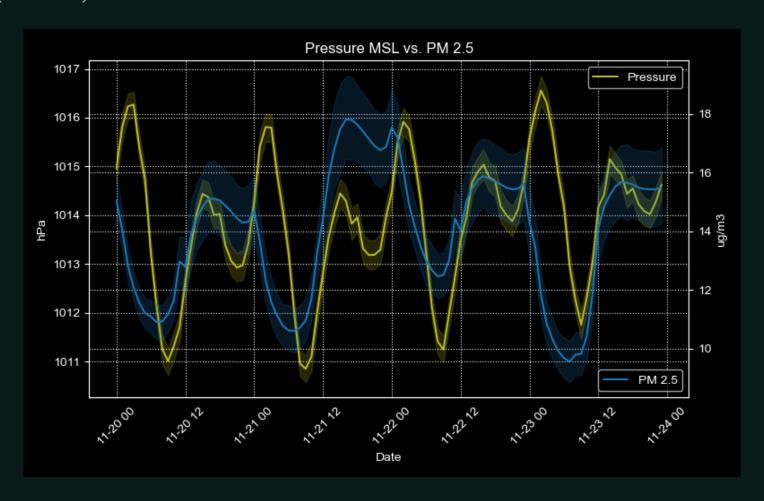




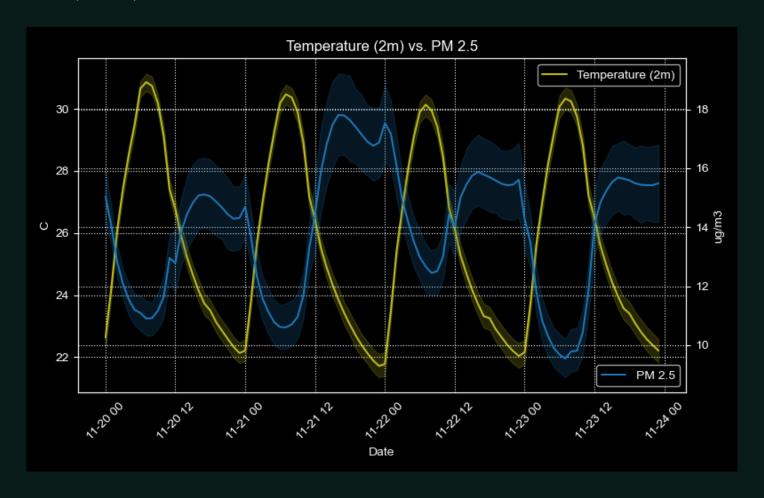
Case: Chiang Mai Hotspot Number



Pressure (MSL) vs PM 2.5



Temperature (2m) vs PM 2.5



Conclusion

- Increased PM2.5 concentrations at night
- PM 2.5 levels are strongly correlated with high atmospheric pressure
- PM 2.5 levels are strongly negative correlated with temperature

Recommendation

- Reduce open burning to mitigate PM2.5
- Stricter Regulations
- Public Awareness Campaigns
- International Cooperation
- Target Audience
 - Government Agencies
 - Media Outlets

Future Works

- Leverage historical data for temporal feature engineering
- Utilize spatial features from adjacent grid areas
- Incorporate historical fire hotspots to predict seasonal fire patterns

Thank you

