

MOTIVATION/INTRODUCTION

WHAT IS THE PROBLEM:

- Goal: predict wildfires in Georgia based on weather forecasts and visualize results in Tableau
- How can we more quickly and accurately predict wildfire occurrences?
- How do we easily display the results to people who may not have a data science background?

WHY IS IT IMPORTANT:

- Wildfire damage accounted for \$18B losses in 2022.
- \$4.4B spent on wildfire suppression in 2021.
- Accurate predictions can guide efficient firefighting resource allocation.
- Communities can be alerted on the event of high risk.
- Potential wins: less casualties, less property damage, preserved ecosystem

OUR APPROACH

WHAT IS THE APPROACH AND HOW DOES IT WORK:

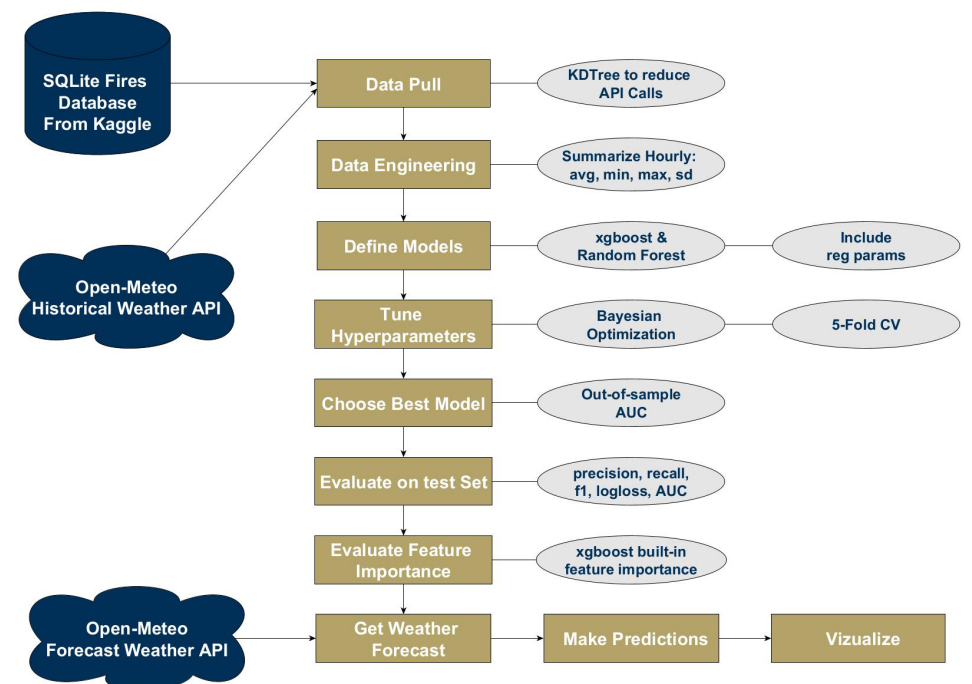
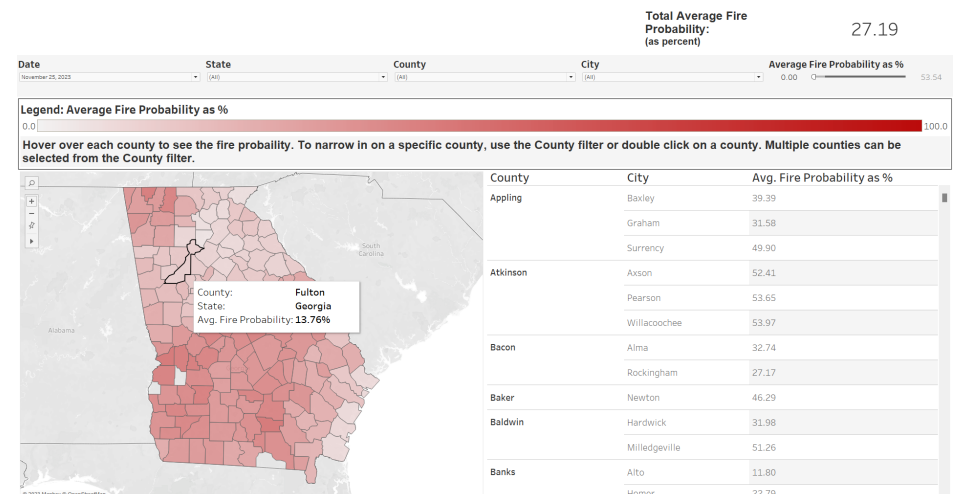
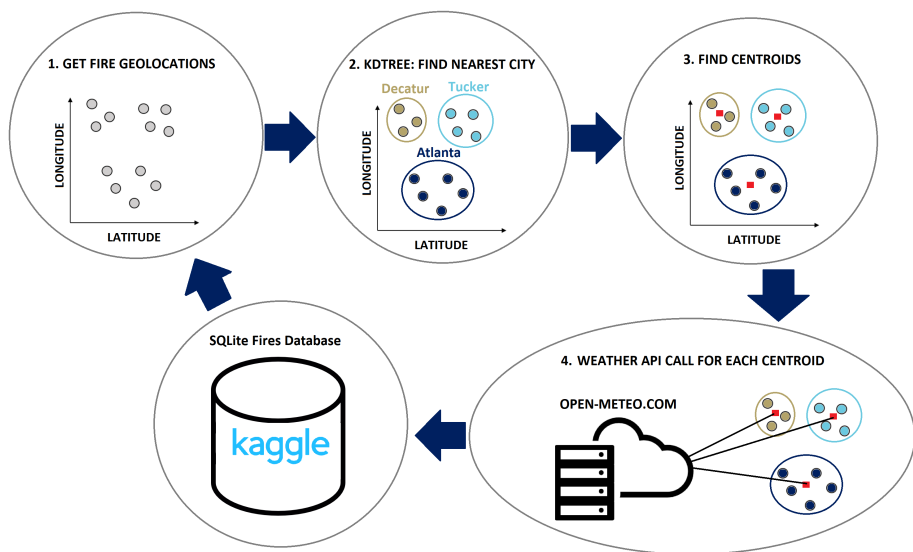
- Choose best between XGboost and Random Forest classification models.
- Elastic net regularization for variable selection.
- Hyperparameter tuning with Bayesian Optimization and k-fold cross-validation.
- Built-in feature importance of xgboost to assess factors of wildfire risk.
- Evaluation metrics on 20% test set: logloss, ROC AUC, precision, recall, f1.
- Visualize the predictions in Tableau.

WHY DOES IT SOLVE THE PROBLEM:

- Computational efficiency of XGBoost and Bayesian Optimization.
- Simplicity and interpretability of Random Forest.
- Cross-validation ensures that model is robust and not prone to overfitting.
- Holistic set of evaluation metrics that indicate the model's real-world relevance.
- Leveraging visualization power of Tableau.

WHAT IS NEW IN OUR APPROACH:

- Robust variable selection through elastic net.
- Capturing non-linear functional forms through ensemble algorithms.
- Enhanced data collection strategy using KDTree.



DATA

HOW DATA WAS COLLECTED:

- Get dataset from Kaggle that contains 1.8 million fire occurrences.
- Identify the nearest city of each fire geo-location using KDTree mapping.
- Reduce geolocations, to a single central location for each city.
- Perform API pulls from open-meteo.com for each of the 660 cities in Georgia.
- Join fire data with weather data and rebalance the dataset.

WHAT ARE THE DATA CHARACTERISTICS:

- Fire data includes: location and time of fire occurrences
- Weather data includes 178 weather features relating to: temperature, humidity, dewpoint, precipitation, snowfall, cloud cover, visibility, evapotranspiration, wind metrics, soil conditions, UV index, and solar radiation.

EXPERIMENTS AND RESULTS

HOW THE PROJECT EVALUATED:

- Model selection based on out-of-sample ROC AUC
- Model accuracy based on 20% test-set: logloss, ROC AUC, Precision, Recall, F1 Score
- Built-in feature importance of xgboost

WHAT ARE THE RESULTS:

- Model is twice better than baseline.
- Precision: 0.55 versus 0.27 baseline.
- Recall: 0.71 versus 0.36 baseline.
- F1 Score: 0.62 versus 0.31 baseline.
- AUC: 0.85 versus 0.5 baseline.
- Logloss: 0.64 versus 0.46 baseline.
- Most important features: relative humidity 2 meters above the ground, temperature and soil-temperature related features, surface pressure, evapotranspiration and terrestrial radiation

HOW OUR METHOD COMPARES TO OTHER METHODS?

Our method is unique in the choice of algorithms, data engineering and data fetching approaches.

