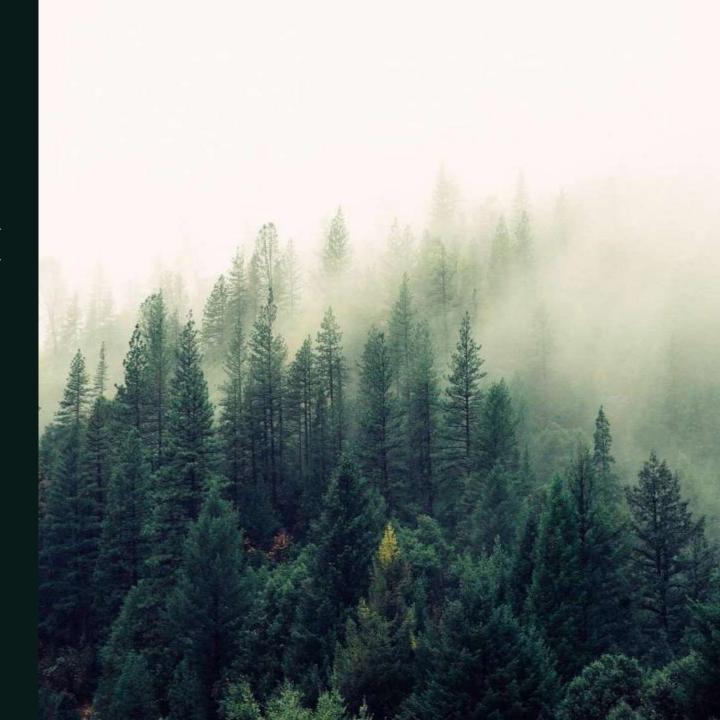
# FOREST FIRE RISK PREDICTION SYSTEM



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## Introduction

Forest fires are severe natural disasters that pose significant threats to the environment, economy and society. Once initiated, they spread rapidly and are challenging to manage effectively. They are often caused by conditions such as high temperatures, low humidity, strong winds, and human activities like deforestation or negligence.

#### Problem Statement

- o Forest fires are highly unpredictable, making early detection and prevention critical.
- o Lack of tools to accurately predict fire-prone areas before disasters occur.
- o Visualizing fire risks geographically can improve resource allocation and response planning.

# Project Overview and Objective

#### Project Goal:

- o Accurately predict forest fire occurrences using historical and environmental data.
- o Identify and visualize fire-prone areas to support proactive disaster management.

#### Key Objectives:

- Utilize advanced machine learning models (Logistic Regression, SVM, Random Forest, Gradient Boosting etc.) for fire prediction.
- o Enhance data by converting spatial coordinates to Latitude and Longitude for better interpretation.
- o Map and visualize fire risk zones over time using tools like Kepler.

## **Dataset Overview**

#### Key Features in the Dataset

X	Y	FFMC	DMC	DC	ISI	Temp	RH	Wind	Rain	y
35.4	-13.5	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	1

Environmental Factors(Features): FFMC, DMC, DC, ISI, Temperature, RH, Wind, Rain.

**Spatial Data:** Coordinates (X, Y) indicating specific locations.

Fire Occurrence (Target Variable):  $y \rightarrow 0$  (no fire) &  $y \rightarrow 1$  (fire)

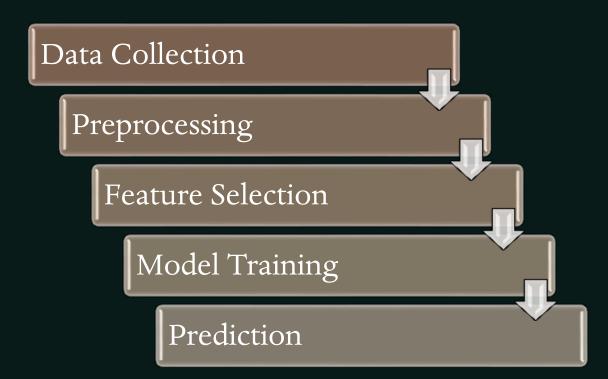
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# Model Implementation and Prediction

# ML Algorithms Used

- o Logistic Regression
- Support Vector Machine (SVM)
- o Random Forest
- Gradient Boosting Classifier

## Implementation



# Geospatial Processing

## Steps

- o Coordinate Scaling
- o Coordinate Conversion to Latitude and Longitude
- o Point ID Assignment

Point ID	Scaled_ Lat	Scaled_ Lon	FFMC	DMC	DC	ISI	Temp	RH	Wind	Rain	y (Fire)
1	42.354	-13.488	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	1

# Visualization with Kepler.gl

#### Why Kepler.gl?

- o Interactive and user-friendly geospatial tool.
- o Showcases fire-prone areas visually on a map.

#### Visualization Example:

- o Red Points: High Fire Risk.
- Yellow Points: No Fire Risk.

## Results

#### Model Accuracy

```
PROBLEMS
           OUTPUT
                    DEBUG CONSOLE
                                   TERMINAL
                                              PORTS
 d----
              09-11-2024
                             22:46
                                                  env
                                                  models
           09-11-2024
                             19:18
 da----
         14-11-2024
                             15:36
                                                  myenv
              21-11-2024
                                                  scripts
                             19:31
PS D:\gda proj> cd scripts
PS D:\gda proj\scripts> python -m venv env
PS D:\gda proj\scripts> .\env\Scripts\activate
(env) PS D:\gda proj\scripts> python working.py
 ---- Logistic Regression -----
 Accuracy: 0.54
 ---- Support Vector Machine ----
 Accuracy: 0.62
 ---- Random Forest Classifier -----
 Accuracy: 0.56
 ---- Gradient Boosting Classifier ----
 Accuracy: 0.53
```

Logistic Regression: 54% Accuracy Support Vector Machine (SVM): 62% Accuracy Random Forest Classifier: 56% Accuracy Gradient Boosting Classifier: 53% Accuracy

## Results

#### Test Case 1: No Fire

Feature	X	Y	FFMC	DMC	DC	ISI	Temperature	RH	Wind	Rain
Value	5	4	91	14.6	25.6	12.3	18	27	5.5	0

PROBLEMS DEBUG CONSOLE TERMINAL PORTS X: 5 Y: 4 FFMC: 91 DMC: 14.6 DC: 25.6 ISI: 12.3 temp: 18 RH: 27 wind: 5.5 rain: 0 Logistic Regression: Prediction: No Fire, Probability of Fire: 0.44, Accuracy: 0.54 Support Vector Machine: Prediction: No Fire, Probability of Fire: 0.45, Accuracy: 0.62 Random Forest Classifier: Prediction: No Fire, Probability of Fire: 0.20, Accuracy: 0.56 Gradient Boosting Classifier: Prediction: No Fire, Probability of Fire: 0.21, Accuracy: 0.53 Do you want to enter another set of values? (yes/no):

## Results

#### Test Case 2: Fire

Feature	X	Y	FFMC	DMC	DC	ISI	Temperature	RH	Wind	Rain
Value	7	5	96.1	181.1	671.2	14.3	21.9	22	10	0

```
PROBLEMS
         OUTPUT DEBUG CONSOLE TERMINAL
                                            PORTS
Gradient Boosting Classifier: Prediction: No Fire, Probability of Fire: 0.21, Accuracy: 0.53
Do you want to enter another set of values? (yes/no): yes
Enter the values for the following features:
X: 7
Y: 5
FFMC: 96.1
DMC: 181.1
DC: 671.2
ISI: 14.3
temp: 21.9
RH: 22
wind: 10
rain: 0
Logistic Regression: Prediction: Fire, Probability of Fire: 0.72, Accuracy: 0.54
Support Vector Machine: Prediction: Fire, Probability of Fire: 0.54, Accuracy: 0.62
Random Forest Classifier: Prediction: Fire, Probability of Fire: 0.64, Accuracy: 0.56
Gradient Boosting Classifier: Prediction: Fire, Probability of Fire: 0.92, Accuracy: 0.53
Do you want to enter another set of values? (yes/no):
```

## Results: Old and new Coordinates

X Y FFMC DMC DC ISI Temp RH Wind Rain Fire Prediction (y)

```
PROBLEMS OUTPUT DEBUG CONSOLE <u>TERMINAL</u> PORTS

Index(['X', 'Y', 'FFMC', 'DMC', 'DC', 'ISI', 'temp', 'RH', 'wind', 'rain', 'y'],
dtype='object')
(env) PS D:\gda_proj\scripts> ■
```

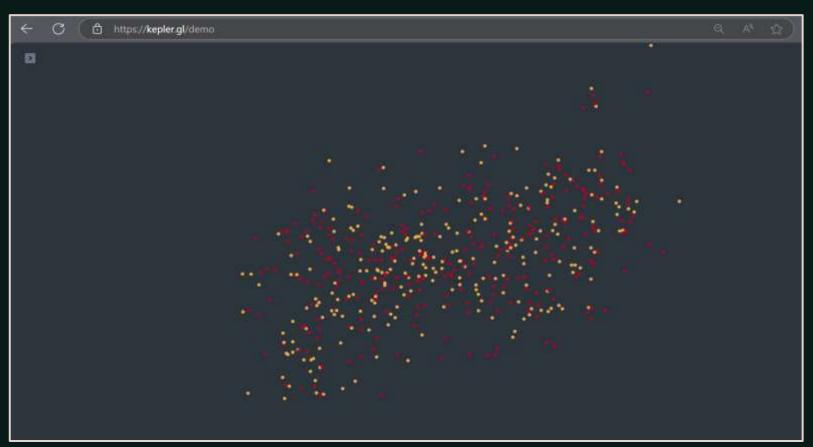
**OLD** 

Scaled	Scaled	Point	FFMC	DMC	DC	ISI	Temp	RH	Wind	Dain	Fire Prediction (y)
Latitude	Longitude	ID	FFIVIC	DMC	DC	101	теттр	КΠ	vv IIIu	Ralli	Fire Prediction (y)

NEW

# Results: Kepler Visualization

## Kepler Visualization of Forest Fire Data Points in Montesinho, Portugal



**Red Points:** Fire Areas

Yellow Points: Non-fire areas

# Future Scope and Conclusion

### Future Scope:

- Enhanced Accuracy: Improve predictions by integrating additional features and exploring deep learning models.
- O **Real-time Application:** Implement live data streams for real-time fire prediction.

#### Conclusion:

- O Successful Prediction: Accurate forest fire forecasting with high precision.
- o Effective Visualization: Actionable insights through Kepler maps for identifying fire-prone areas.

# Thank you