

File Edit Selection View Help PythonProject_1.ipynb - Jupyter Studio Code

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PythonProject_1.ipynb X

C: > Users > USER > Downloads > PythonProject_1.ipynb > ##Earthquake Prediction

+ Code + Markdown ...

Select Kernel

#Earthquake Prediction Gayathry M Wariyar
AM.EN.U4AIE21130

```
import numpy as np
import pandas as pd
import requests
from sklearn import preprocessing
import matplotlib.pyplot as plt
import seaborn as sns
from pandas.plotting import scatter_matrix
from sklearn.impute import SimpleImputer
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
import time
```

[398] Python

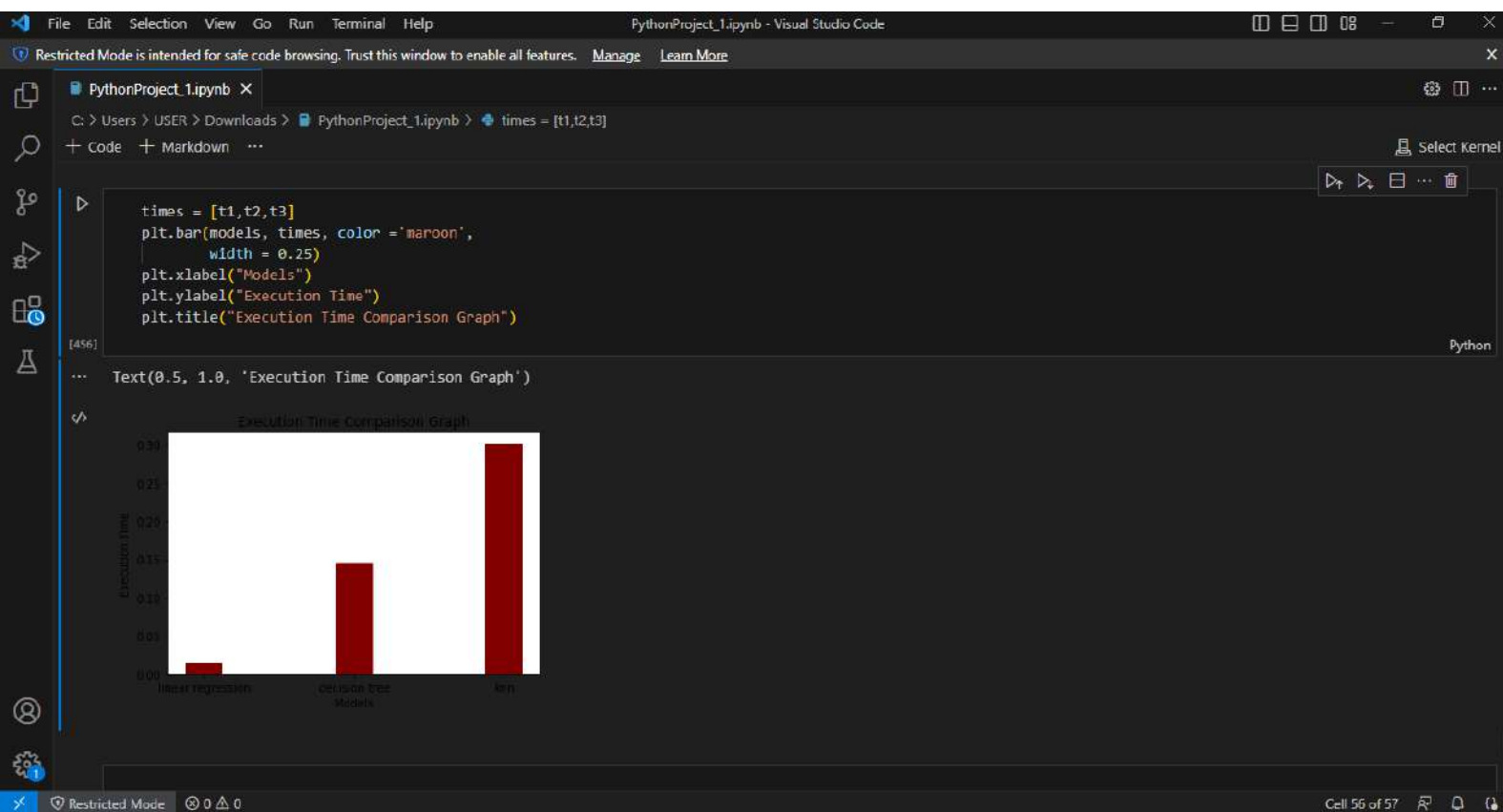
```
from google.colab import drive
drive.mount('/content/drive')
df = pd.read_csv("/content/drive/MyDrive/Colab Notebooks/earthquake_prediction/earthquake1.csv")
```

[399] Python

... Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
df.info()
```

Cell 1 of 57



C:\Users\USER\Downloads> PythonProject_1.ipynb > plt.plot(y_test, ans1, 'o')

+ Code + Markdown ...

Select Kernel

MEAN ABSOLUTE ERROR: 0.05878240403205080

Mean Squared Error: 0.00625827169726636

Root Mean Squared Error: 0.07910923901331854

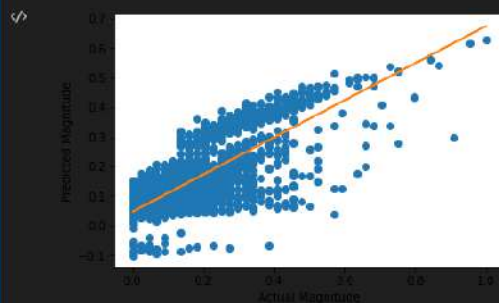


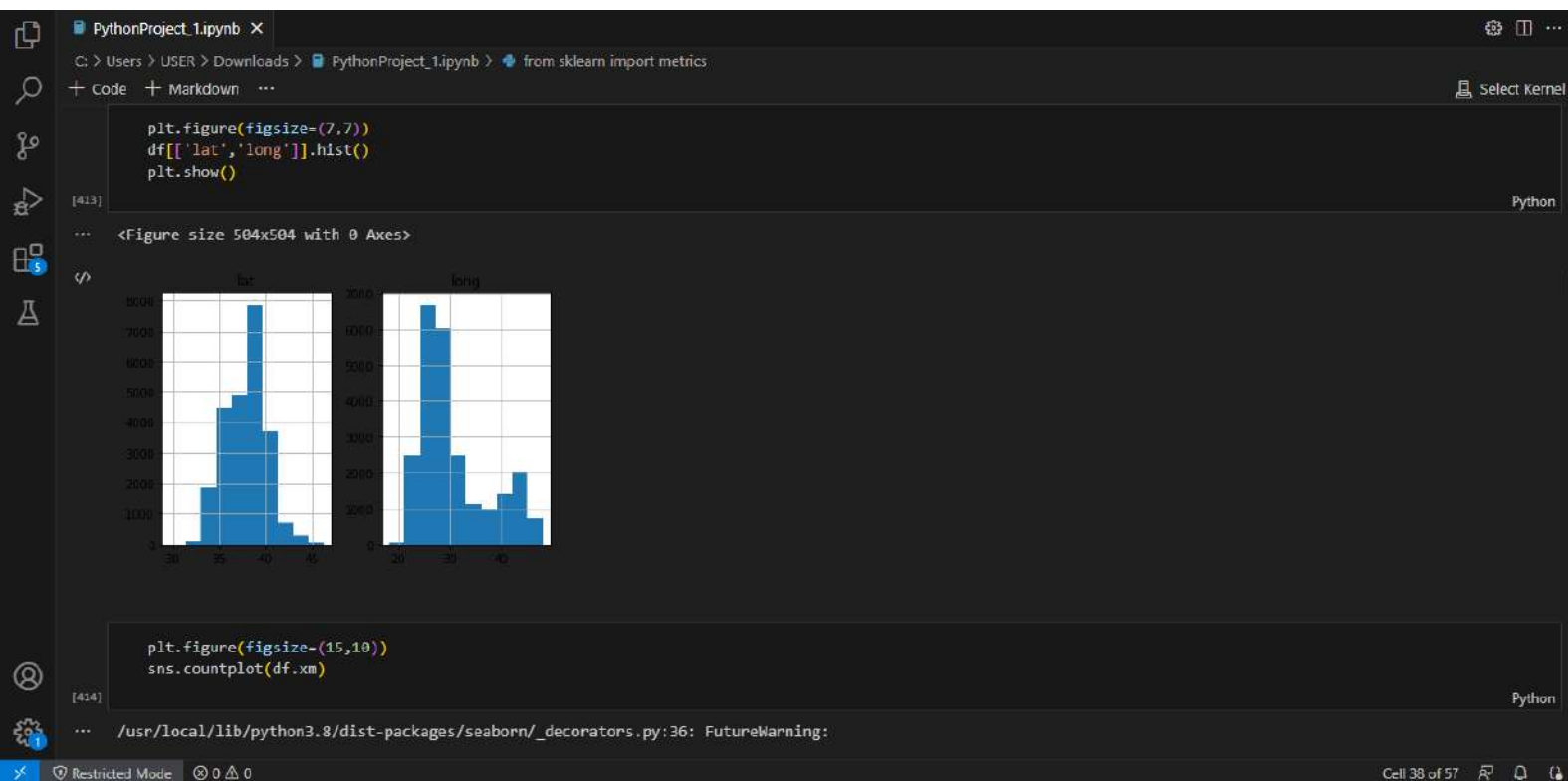
```
plt.plot(y_test, ans1, 'o')
m, b = np.polyfit(y_test, ans1, 1)
plt.plot(y_test, m*y_test + b)
plt.xlabel("Actual Magnitude")
plt.ylabel("Predicted Magnitude")
```

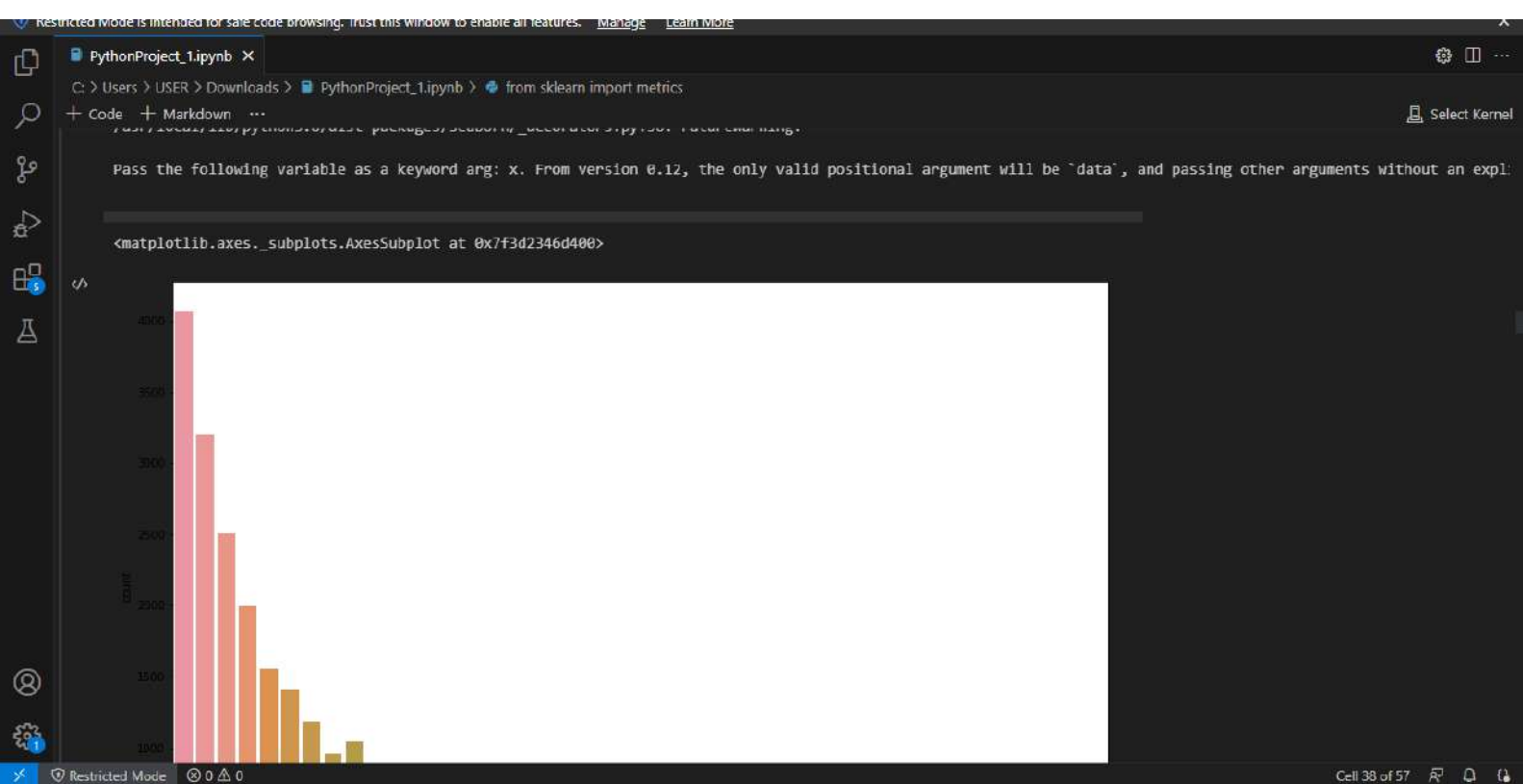
[428]

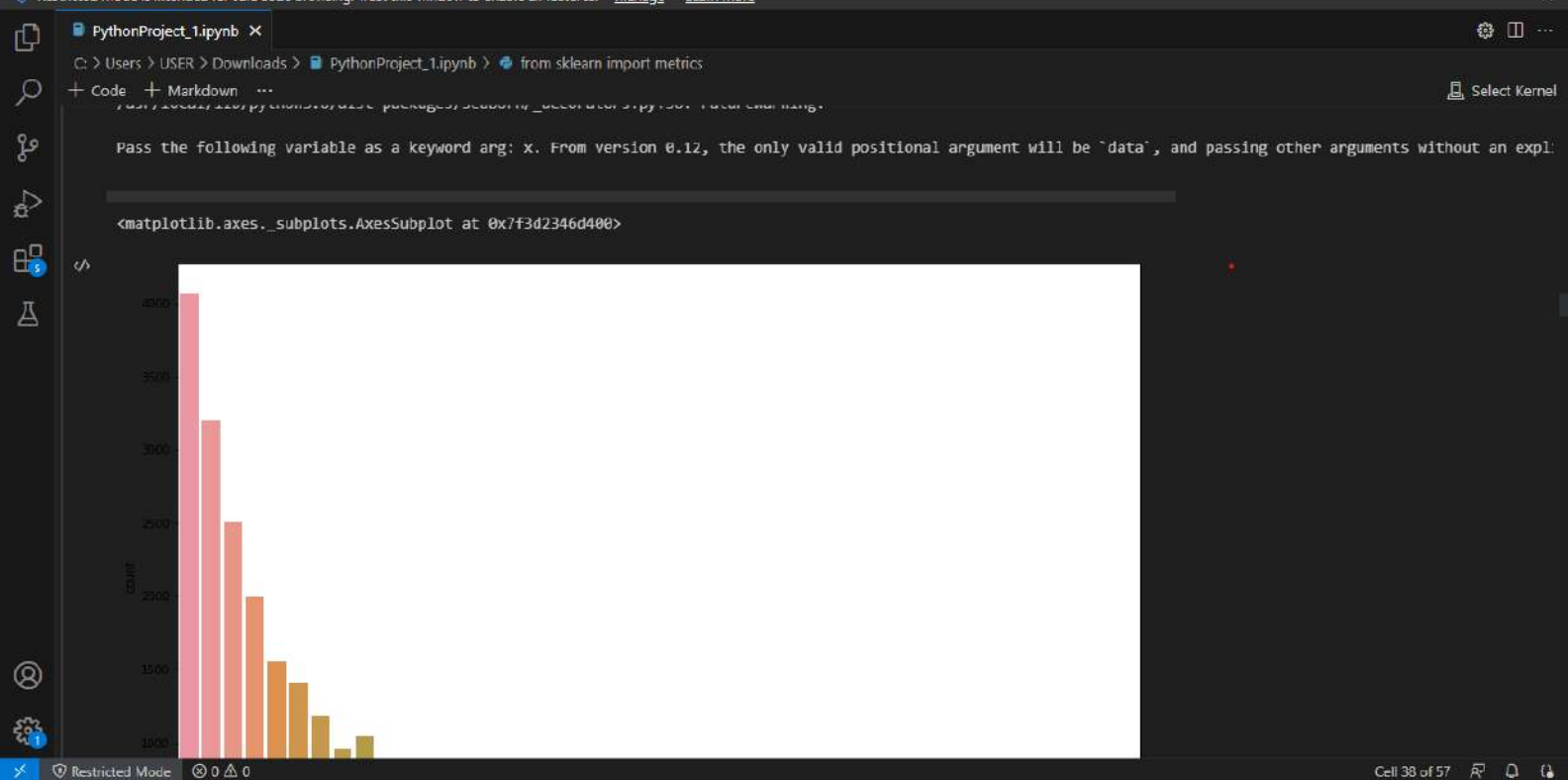
Python

```
... Text(0, 0.5, 'Predicted Magnitude')
```



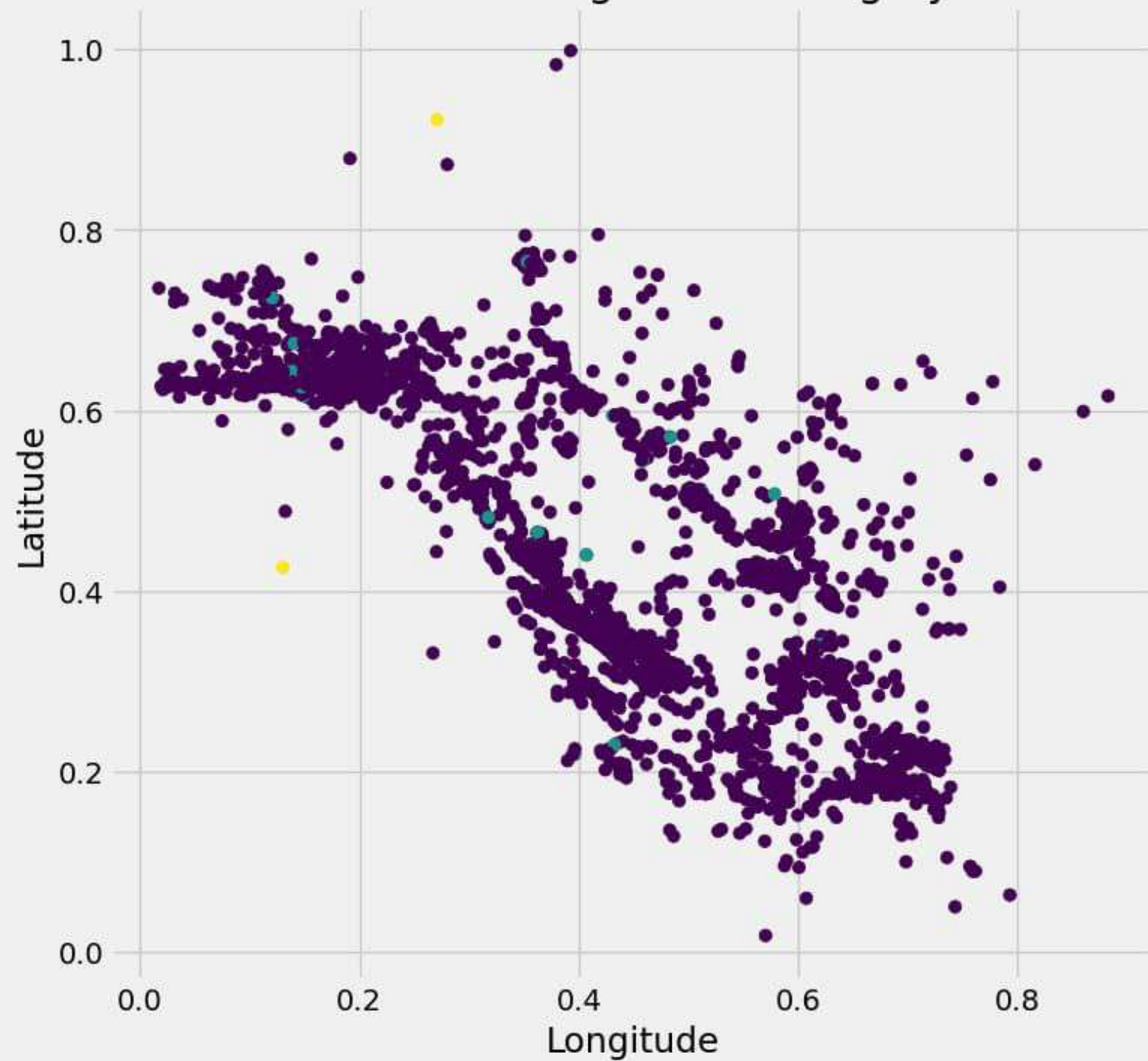


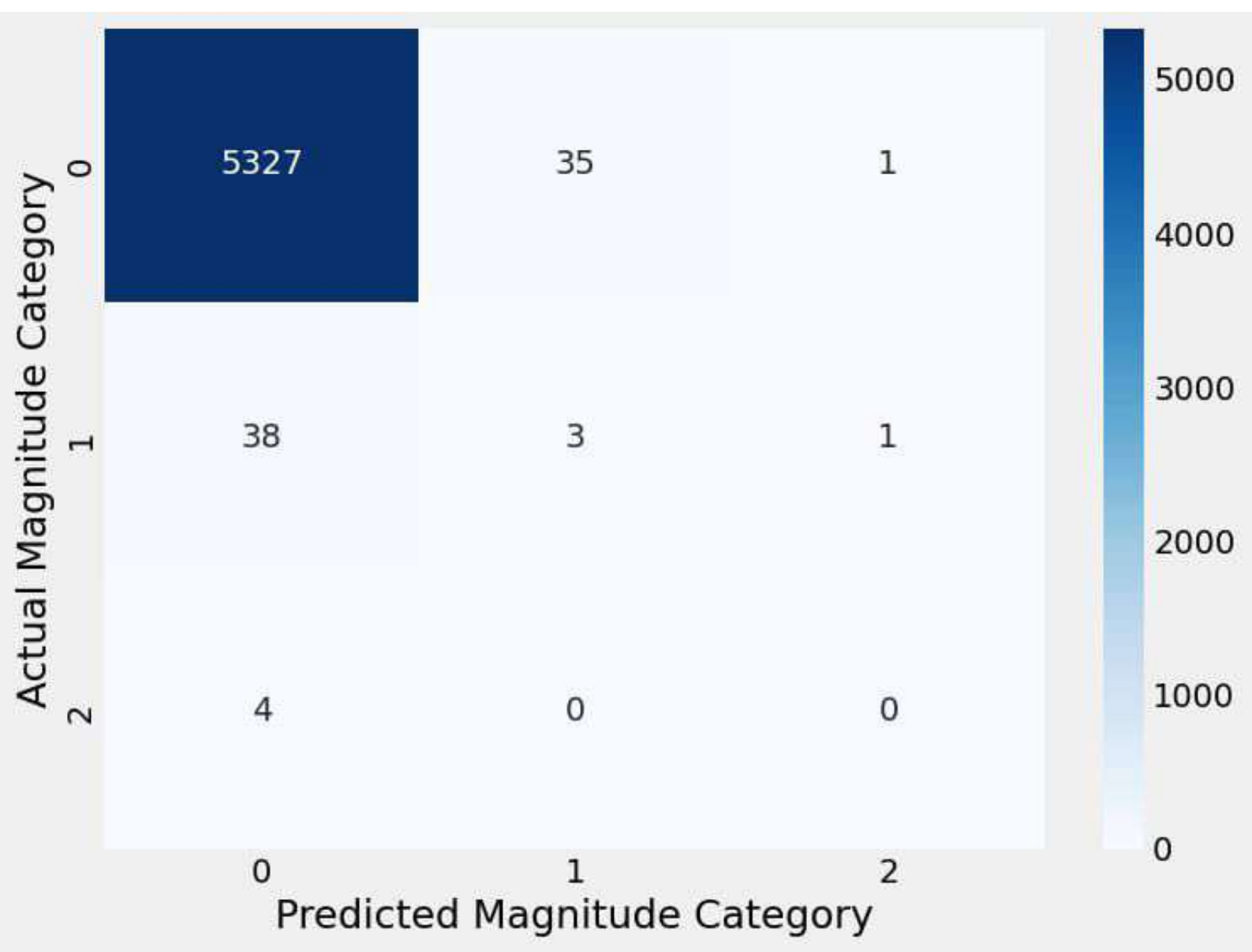




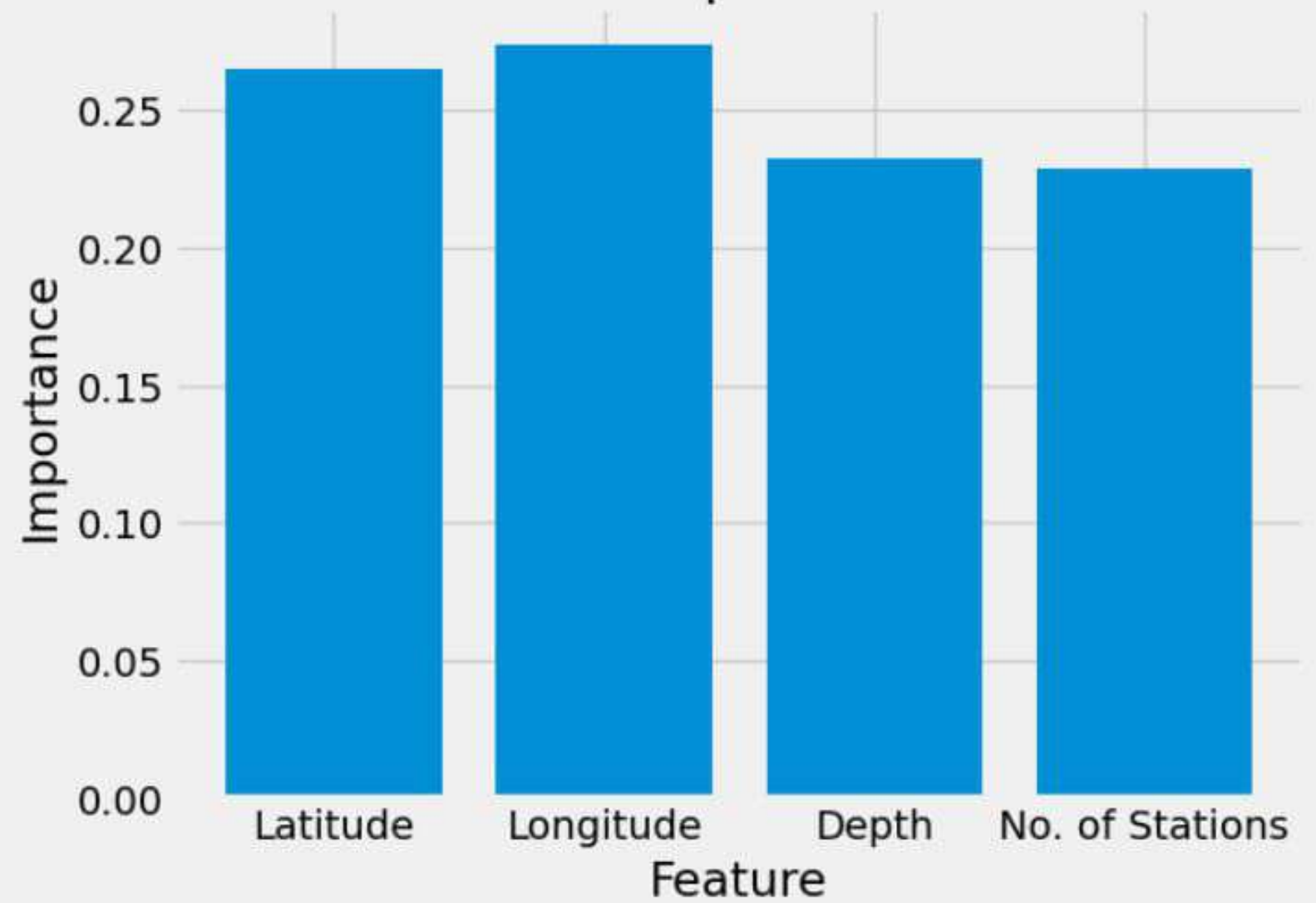


Predicted Magnitude Category





Feature Importance Plot



V. EVALUATION AND DISCUSSIONS

DATA PRE-PROCESSING

For both the datasets the following pre-processing steps were taken:

- First, we dropped the 'id' column since it is not useful for prediction.

```
df = df.drop('id',axis=1)
```

- Next we converted categorical variables into numerical values so that it could be easily fitted to a machine learning model using a label encoder.

```
# Data Encoding
label_encoder = preprocessing.LabelEncoder()
for col in df.columns:
    if df[col].dtype == 'object':
        label_encoder.fit(df[col])
        df[col] = label_encoder.transform(df[col])
df.dtypes
```

- Imputation with SimpleImputer which replaced the missing data with mean was done to handle null values.

```
# Imputing Missing Values with Mean
si=SimpleImputer(missing_values = np.nan, strategy="mean")
si.fit(df[["dist","mw"]])
df[["dist","mw"]] = si.transform(df[["dist","mw"]])
df.isnull().sum()
```

- Normalization of the data was done using MinMax

```
import datetime
import time

timestamp = []
for d, t in zip(df['date'], df['time']):
    ts = datetime.datetime.strptime(d+' '+t, '%Y.%m.%d %I:%M:%S %p')
    timestamp.append(time.mktime(ts.timetuple()))
timestamp = pd.Series(timestamp)
df['Timestamp'] = timestamp.values
final_data = df.drop(['date', 'time'], axis=1)
final_data = final_data[final_data.Timestamp != 'ValueError']
df = final_data
df.head()
```

SPLITTING THE DATASET

The datasets are split into training and testing datasets.

The dataset 1 is split into 80% training and 20% testing, with a random state of 2, meaning that the same random sample will be used each time the code is run.

```
y=np.array(df['xm'])
X=np.array(df.drop('xm',axis=1))

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test
size=0.2,random state=2)
```

The dataset 2 is split into 75% and 25% for training and testing respectively with a random state of 2.

```
y=np.array(df['xm'])
X=np.array(df.drop('xm',axis=1))
```

B. Decision Tree

```
from sklearn.tree import DecisionTreeRegressor
start2 = time.time()
regressor = DecisionTreeRegressor(random_state = 40)
regressor.fit(X_train,y_train)
ans2 = regressor.predict(X_test)
end2 = time.time()
t2 = end2-start2
```

C. KNN

```
from sklearn.neighbors import KNeighborsRegressor
start3 = time.time()
knn = KNeighborsRegressor(n_neighbors=6)
knn.fit(X_train, y_train)
ans3 = knn.predict(X_test)
end3 = time.time()
t3 = end3 - start3
```

RESULTS

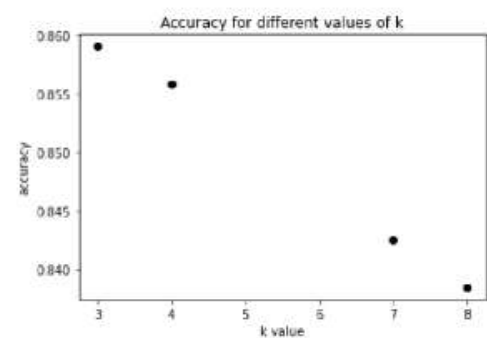
DATASET 1: earthquake1.csv

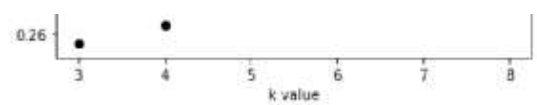
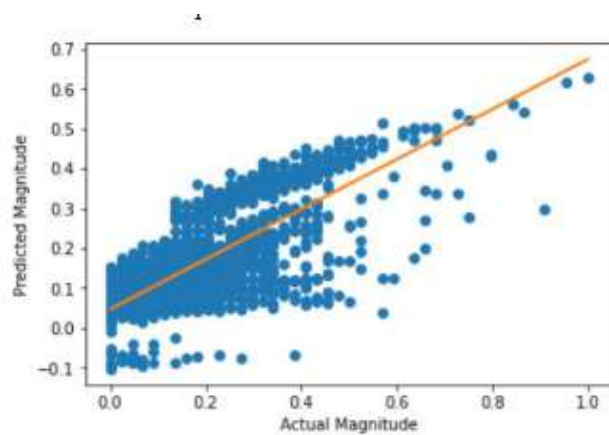
Linear Regression Model

- Accuracy of Linear Regression model is: 0.63134131503029
- Mean Absolute Error: 0.05878246463205686
- Mean Squared Error: 0.00625827169726636
- Root Mean Squared Error: 0.07910923001331854

KNN Model

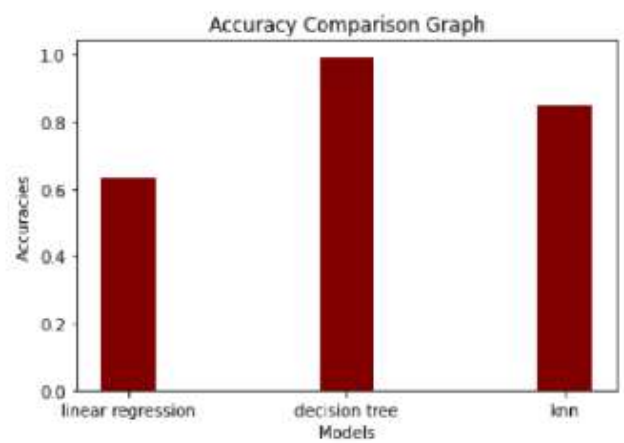
- Accuracy of KNN model is: 0.8457466919393031
- Mean Absolute Error: 0.03305598677318794
- Mean Squared Error: 0.002618571462992348
- Root Mean Squared Error: 0.051171979275696854

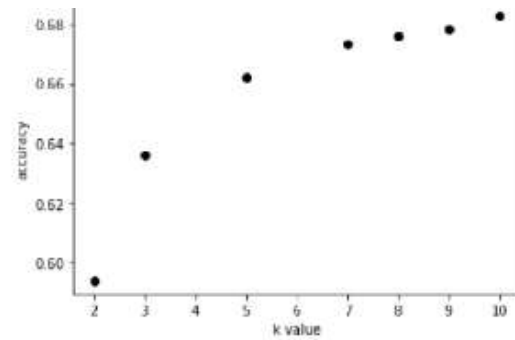
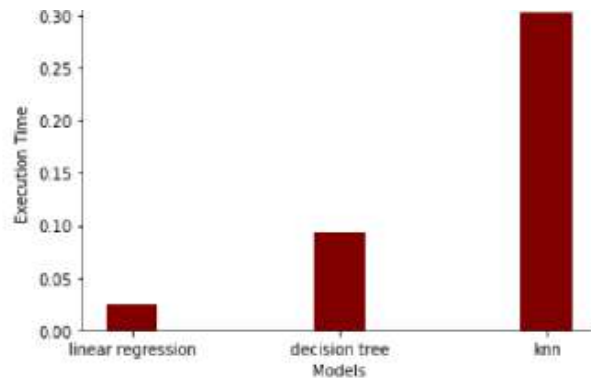




Decision Tree Model

- Accuracy of Decision Tree model is: 0.9932960893884235.
- Mean Absolute Error: 0.0006909999621372331
- Mean Squared Error: 0.00011380416561969702
- Root Mean Squared Error: 0.010667903525046383





DATASET 2: earthquake2.csv

Linear Regression Model

- Accuracy of Linear Regression model is: 0.6520026760336074
- Mean Absolute Error: 0.0447298826759214
- Mean Squared Error: 0.0034475163509273773
- Root Mean Squared Error: 0.058715554590988726

