



4222-SURYA GROUP OF INSTITUTIONS

VIKARAVANDI -605 652



PROJECT NAME:

EARTHQUAKE-PREDICTION-USING-PYTHON

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ECE DEPARTMENT

EARTHQUAKE-PREDICTION-USING-PYTHON

INTRODUCTION:

Earthquake prediction is a challenging and complex task that is still an active area of research. It is a way to predict the magnitude of earthquake based on parameters such as longitude, latitude, depth, and duration magnitude, country. These approaches are based on the analysis of seismic data, historical earthquake data, and other relevant factors. People used to minimize loss of life and property.

ML MODELS USED:

- Linear Regression
- Decision Tree
- K-Nearest Neighbors

STEPS TAKEN:

- Data source
- Feature exploration
- Visualization
- Data splitting
- Training and evaluation

DATA SOUCE:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
import os
print(os.listdir("../input"))
```

['database.csv']

| SI NO | Date | Time | Latitude | Longitude | Depth | Magnitude |
|-------|------------|----------|----------|-----------|-------|-----------|
| 0 | 01/02/1965 | 13:44:18 | 19.246 | 145.616 | 131.6 | 6.0 |
| 1 | 01/04/1965 | 11:29:49 | 1.863 | 127.352 | 80.0 | 5.8 |
| 2 | 01/05/1965 | 18:05:58 | -20.579 | -173.972 | 20.0 | 6.2 |
| 3 | 01/08/1965 | 18:49:43 | -59.076 | -23.557 | 15.0 | 5.8 |
| 4 | 01/09/1965 | 13:32:50 | 11.938 | 126.427 | 15.0 | 5.8 |

FEATURE EXPLORATION:

```
Index(['Date', 'Time', 'Latitude', 'Longitude', 'Type', 'Depth', 'Depth Error',  
      'Depth Seismic Stations', 'Magnitude', 'Magnitude Type',  
      'Magnitude Error', 'Magnitude Seismic Stations', 'Azimuthal Gap',  
      'Horizontal Distance', 'Horizontal Error', 'Root Mean Square', 'ID',  
      'Source', 'Location Source', 'Magnitude Source', 'Status'],  
      dtype='object')
```

Figure out the main features from earthquake data and create a object of that features, namely, Date, Time, Latitude, Longitude, Depth, Magnitude.

```
data = data[['Date', 'Time', 'Latitude', 'Longitude', 'Depth', 'Magnitude']]  
data.head()
```

Out[4]:

| | Date | Time | Latitude | Longitude | Depth | Magnitude |
|---|------------|----------|----------|-----------|-------|-----------|
| 0 | 01/02/1965 | 13:44:18 | 19.246 | 145.616 | 131.6 | 6.0 |
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Visualization:

Here, all the earthquakes from the database in visualized on to the world map which shows clear representation of the locations where frequency of the earthquake will be more.

In [8]:

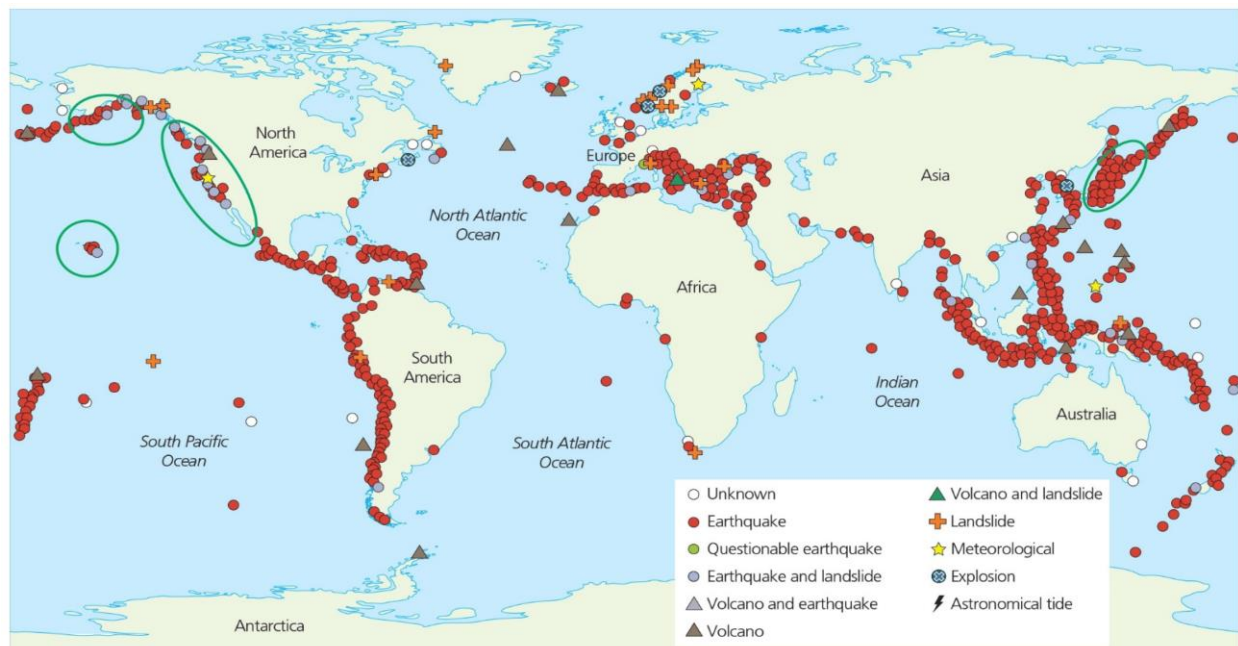
```
from mpl_toolkits.basemap import Basemap
```

```
m = Basemap(projection='mill',llcrnrlat=-80,urcnrlat=80,llcrnrlon=-  
180,urcnrlon=180,lat_ts=20,resolution='c')
```

```

longitudes = data["Longitude"].tolist()
latitudes = data["Latitude"].tolist()
#m = Basemap(width=12000000,height=9000000,projection='lcc',
             #resolution=None,lat_1=80.,lat_2=55,lat_0=80,lon_0=-107.)
x,y = m(longitudes,latitudes)
In [9]:
fig = plt.figure(figsize=(12,10))
plt.title("All affected areas")
m.plot(x, y, "o", markersize = 2, color = 'blue')
m.drawcoastlines()
m.fillcontinents(color='coral',lake_color='aqua')
m.drawmapboundary()
m.drawcountries()
plt.show()

```



Data splitting:

The data split was 90% train and 10% test.

| Weekly model | | | | Daily model | | | |
|--------------|--------------|---------|--------|---------------|---------|--------|--|
| | Records | Balance | Events | Records | Balance | Events | |
| Train | 95,181 (90%) | 6.95% | 6,612 | 666,688 (90%) | 1.72% | 11,450 | |
| Test | 11,084 (10%) | 8.46% | 938 | 77,606 (10%) | 2.16% | 1,677 | |

TRAINING AND EVALUATION

demonstrate that the train-test split procedure is repeatable

```
from sklearn.datasets import make_blobs
from sklearn.model_selection import train_test_split
# create dataset
X, y = make_blobs(n_samples=100)
# split into train test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=1)
# summarize first 5 rows
print(X_train[:5, :])
# split again, and we should see the same split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=1)
# summarize first 5 rows
print(X_train[:5, :])
```

```
[[-2.54341511  4.98947608]
 [ 5.65996724 -8.50997751]
 [-2.5072835  10.06155749]
 [ 6.92679558 -5.91095498]
 [ 6.01313957 -7.7749444 ]]
```

```
[[-2.54341511  4.98947608]
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