



4222-SURYA GROUP OF INSTITUTIONS VIKRAVANDI-605652

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',

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
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
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Visualization:

Here, all the earthquakes from the database in visualized on to the world map which shows clear

^{&#}x27;Depth Seismic Stations', 'Magnitude', 'Magnitude Type',

^{&#}x27;Magnitude Error', 'Magnitude Seismic Stations', 'Azimuthal Gap',

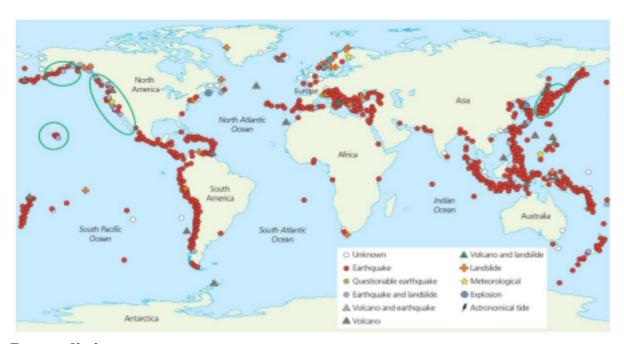
^{&#}x27;Horizontal Distance', 'Horizontal Error', 'Root Mean Square', 'ID',

^{&#}x27;Source', 'Location Source', 'Magnitude Source', 'Status'],

In [8]:

from mpl_toolkits.basemap import Basemap

```
m = Basemap(projection='mill',llcrnrlat=-80,urcrnrlat=80, llcrnrlon=-
180,urcrnrlon=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]:
<u>fig</u> = plt.<u>figure</u>(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

rain	Records	Balance	Events	
	95,181 (98%)	6.95%	6,612	
est	11,084 (18%)	8.46%	938	

Daily model

Records	Balance	Events
666,688 (90%)	1.72%	11,450
77,686 (18%)	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]
[5.65996724 -8.50997751]
[-2.5072835 10.06155749]
[6.92679558 -5.91095498]
[6.01313957 -7.7749444 ]]
```