

# Phase 5: Project Documentation & Submission

## 1.1 Understanding the problem :

The problem is to develop an earthquake prediction model using a Kaggle dataset. The objective is to explore and understand the key features of earthquake data, visualize the data on a world map for a global overview, split the data for training and testing, and build a neural network model to predict earthquake magnitudes based on the given features.

## 2.1 Design Thinking :

Our approach to solve this problem involves a structured design thinking process, ensuring systematic and effective analysis of Earthquake prediction.

## 3.1 Data Collection:

For this project, we use a dataset from kaggle which is recommended for our problem statement. The link of the dataset is given below .

**Dataset Link:** <https://www.kaggle.com/datasets/usgs/earthquake-database>

## 4.1 Code Implementation:

```
import numpy as np
import pandas as pd
from keras.models import Sequential
from keras.layers import Dense
import folium

# Predicting the earthquake intensity which is given by the user.
test_longitude = float(input("Enter longitude : "))    # test value : -118.0
test_latitude = float(input("Enter latitude : "))      # test value : 34.0

# Dataset recommended by Naan Mudhalvan
data = pd.read_csv('/kaggle/input/earthquake-database/database.csv')
```

```
df = pd.DataFrame(data)
```

#### 4.1.1 Normalization

As mentioned before, our model requires latitude and longitude as input and intensity, lat and long of earthquake for processing. The dataset given for us already contains latitude and longitude, so we are not modifying any fields of the dataset.

```
# Normalize longitude and latitude
```

```
df['normalized_longitude'] = (df['Longitude'] - df['Longitude'].mean()) /  
df['Longitude'].std()
```

```
df['normalized_latitude'] = (df['Latitude'] - df['Latitude'].mean()) / df['Latitude'].std()
```

```
# Scale earthquake intensity between 0 and 1
```

```
df['scaled_intensity'] = (df['Magnitude'] - df['Magnitude'].min()) / (df['Magnitude'].max() -  
df['Magnitude'].min())
```

#### 4.1.2 Creating a Neural Network

```
# Create a neural network model for regression
```

```
model = Sequential()  
model.add(Dense(units=32, activation='relu', input_dim=2))  
model.add(Dense(units=1, activation='sigmoid'))
```

```
# Compile the model
```

```
model.compile(loss='mean_squared_error', optimizer='adam')
```

```
# Train the model using normalized longitude, latitude, and scaled intensity
```

```
model.fit(df[['normalized_longitude', 'normalized_latitude']], df['scaled_intensity'],  
epochs=10, batch_size=32)
```

```
# Normalize the test longitude and latitude
```

```
normalized_test_longitude = (test_longitude - df['Longitude'].mean()) /  
df['Longitude'].std()  
normalized_test_latitude = (test_latitude - df['Latitude'].mean()) / df['Latitude'].std()
```

```
# Predict the earthquake intensity for the test location
```

```
predicted_intensity = model.predict(np.array([[normalized_test_longitude,  
normalized_test_latitude]]))[0][0]
```

### 4.1.3 Visualization

To effectively display our findings we will be displaying the output of our model in a map. For visualization we will be using folium library from python in order to display our output in map.

```
# Create a base map
map = folium.Map(location=[test_latitude, test_longitude], zoom_start=12)

# Adding a circle around intensity area
folium.CircleMarker(
    location=[test_latitude, test_longitude],
    radius= predicted_intensity * 20,
    color='red',
    fill=True,
    fill_color='red',
    fill_opacity=0.6
).add_to(map)

#Saving the map as intensity_map.html
map.save('intensity_map.html')
```

### 4.1.4 Output:

Enter longitude : 80.270721

Enter latitude : 13.082680

Epoch 1/10

732/732 [=====] - 1s 1ms/step - loss: 0.0370

Epoch 2/10

732/732 [=====] - 1s 1ms/step - loss: 0.0141

Epoch 3/10

732/732 [=====] - 1s 1ms/step - loss: 0.0139

Epoch 4/10

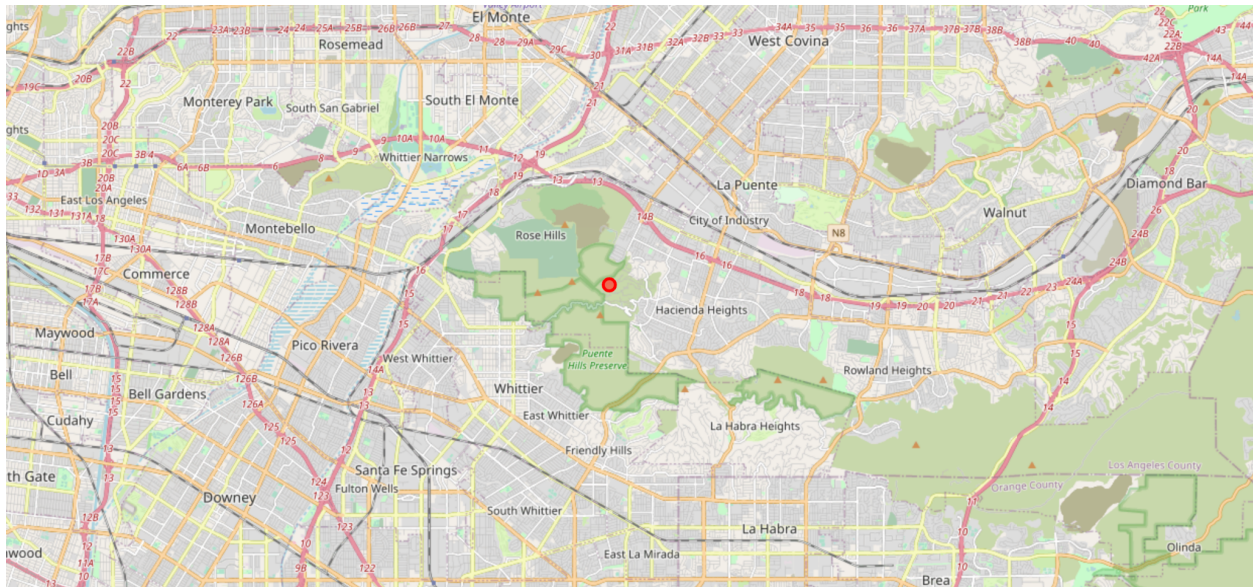
732/732 [=====] - 1s 1ms/step - loss: 0.0138

Epoch 5/10

732/732 [=====] - 1s 1ms/step - loss: 0.0138



Intensity\_map.html



The red dot represents the intensity of earthquakes in the given region. Wider the dot more effective the earthquake will be.