

Importing libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

Importing dataset

```
In [2]: data=pd.read_csv(r"C:\Users\user\Downloads\cities.csv")
data
```

```
Out[2]:
```

| | id | name | state_id | state_code | state_name | country_id | country_code | country_name |
|--------|--------|---------------------|----------|------------|-------------------|------------|--------------|--------------|
| 0 | 52 | Ashkāsham | 3901 | BDS | Badakhshan | 1 | AF | Afghanistan |
| 1 | 68 | Fayzabad | 3901 | BDS | Badakhshan | 1 | AF | Afghanistan |
| 2 | 78 | Jurm | 3901 | BDS | Badakhshan | 1 | AF | Afghanistan |
| 3 | 84 | Khandūd | 3901 | BDS | Badakhshan | 1 | AF | Afghanistan |
| 4 | 115 | Rāghistān | 3901 | BDS | Badakhshan | 1 | AF | Afghanistan |
| ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 150449 | 131496 | Redcliff | 1957 | MI | Midlands Province | 247 | ZW | Zimbabwe |
| 150450 | 131502 | Shangani | 1957 | MI | Midlands Province | 247 | ZW | Zimbabwe |
| 150451 | 131503 | Shurugwi | 1957 | MI | Midlands Province | 247 | ZW | Zimbabwe |
| 150452 | 131504 | Shurugwi District | 1957 | MI | Midlands Province | 247 | ZW | Zimbabwe |
| 150453 | 131508 | Zvishavane District | 1957 | MI | Midlands Province | 247 | ZW | Zimbabwe |

150454 rows × 11 columns



info

```
In [3]: # to identify missing values
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150454 entries, 0 to 150453
Data columns (total 11 columns):
#   Column          Non-Null Count  Dtype
---  -
#   Column          Non-Null Count  Dtype
```

```
0 id 150454 non-null int64
1 name 150454 non-null object
2 state_id 150454 non-null int64
3 state_code 150129 non-null object
4 state_name 150454 non-null object
5 country_id 150454 non-null int64
6 country_code 150406 non-null object
7 country_name 150454 non-null object
8 latitude 150454 non-null float64
9 longitude 150454 non-null float64
10 wikiDataId 147198 non-null object
dtypes: float64(2), int64(3), object(6)
memory usage: 12.6+ MB
```

describe

In [4]:

```
# to display summary of the dataset
data.describe()
```

Out[4]:

| | id | state_id | country_id | latitude | longitude |
|-------|---------------|---------------|---------------|---------------|---------------|
| count | 150454.000000 | 150454.000000 | 150454.000000 | 150454.000000 | 150454.000000 |
| mean | 76407.091689 | 2678.377677 | 140.658460 | 31.556175 | 2.369557 |
| std | 44357.755335 | 1363.513591 | 70.666123 | 22.813220 | 68.012770 |
| min | 1.000000 | 1.000000 | 1.000000 | -75.000000 | -179.121980 |
| 25% | 38160.250000 | 1451.000000 | 82.000000 | 19.000000 | -58.468150 |
| 50% | 75975.500000 | 2174.000000 | 142.000000 | 40.684720 | 8.669980 |
| 75% | 115204.750000 | 3905.000000 | 207.000000 | 47.239220 | 27.750000 |
| max | 153528.000000 | 5116.000000 | 247.000000 | 73.508190 | 179.466000 |

columns

In [5]:

```
# to display headings of the dataset
data.columns
```

Out[5]:

```
Index(['id', 'name', 'state_id', 'state_code', 'state_name', 'country_id',
      'country_code', 'country_name', 'latitude', 'longitude', 'wikiDataId'],
      dtype='object')
```

In [6]:

```
a=data.dropna(axis=1)
a
```

Out[6]:

| | id | name | state_id | state_name | country_id | country_name | latitude | longitude |
|---|----|-----------|----------|------------|------------|--------------|----------|-----------|
| 0 | 52 | Ashkāsham | 3901 | Badakhshan | 1 | Afghanistan | 36.68333 | 71.53333 |
| 1 | 68 | Fayzabad | 3901 | Badakhshan | 1 | Afghanistan | 37.11664 | 70.58002 |
| 2 | 78 | Jurm | 3901 | Badakhshan | 1 | Afghanistan | 36.86477 | 70.83421 |

| | id | name | state_id | state_name | country_id | country_name | latitude | longitude |
|---------------|--------|---------------------|----------|-------------------|------------|--------------|-----------|-----------|
| 3 | 84 | Khandūd | 3901 | Badakhshan | 1 | Afghanistan | 36.95127 | 72.31800 |
| 4 | 115 | Rāghistān | 3901 | Badakhshan | 1 | Afghanistan | 37.66079 | 70.67346 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 150449 | 131496 | Redcliff | 1957 | Midlands Province | 247 | Zimbabwe | -19.03333 | 29.78333 |
| 150450 | 131502 | Shangani | 1957 | Midlands Province | 247 | Zimbabwe | -19.78333 | 29.36667 |
| 150451 | 131503 | Shurugwi | 1957 | Midlands Province | 247 | Zimbabwe | -19.67016 | 30.00589 |
| 150452 | 131504 | Shurugwi District | 1957 | Midlands Province | 247 | Zimbabwe | -19.75000 | 30.16667 |
| 150453 | 131508 | Zvishavane District | 1957 | Midlands Province | 247 | Zimbabwe | -20.30345 | 30.07514 |

150454 rows × 8 columns

In [7]:

```
a.columns
```

Out[7]: Index(['id', 'name', 'state_id', 'state_name', 'country_id', 'country_name', 'latitude', 'longitude'], dtype='object')

To train the model-Model Building

In [8]:

```
x=a[['id']]
y=a['country_id']
```

In [9]:

```
# to split my dataset into training and test data
from sklearn.model_selection import train_test_split

x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

Linear regression

In [10]:

```
from sklearn.linear_model import LinearRegression
lr= LinearRegression()
lr.fit(x_train,y_train)
```

Out[10]: LinearRegression()

In [11]:

```
print(lr.intercept_)
```

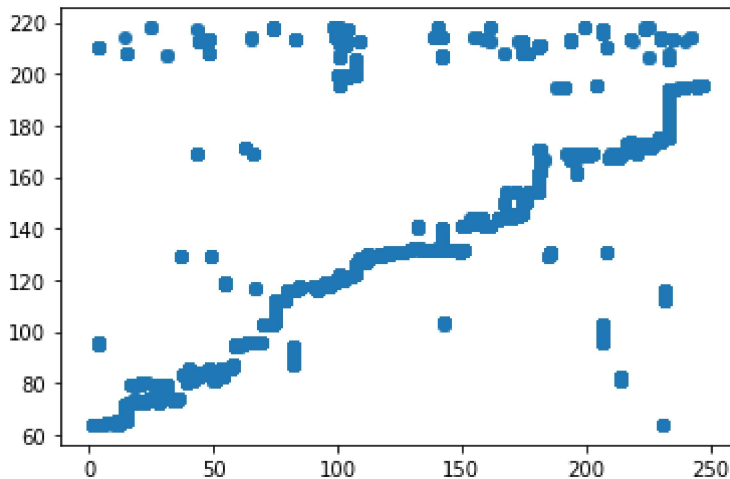
63.976158623252175

```
In [12]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

```
Out[12]:      Co-efficient
id      0.001002
```

```
In [13]: prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

```
Out[13]: <matplotlib.collections.PathCollection at 0x1e60cbd11f0>
```



```
In [14]: print(lr.score(x_test,y_test))

0.4002211834789693
```

```
In [15]: lr.score(x_train,y_train)
```

```
Out[15]: 0.39602347546627725
```

Ridge regression

```
In [16]: from sklearn.linear_model import Ridge,Lasso
```

```
In [17]: rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
rr.score(x_test,y_test)
```

```
Out[17]: 0.4002211834789692
```

```
In [18]: rr.score(x_train,y_train)
```

Out[18]: 0.39602347546627725

Lasso regression

```
In [19]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
         la.score(x_train,y_train)
```

Out[19]: 0.3960234754561025

```
In [20]: la.score(x_test,y_test)
```

Out[20]: 0.4002211608355648

Elastic net regression

```
In [21]: from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
```

Out[21]: ElasticNet()

```
In [22]: print(en.coef_)
```

[0.00100244]

```
In [23]: print(en.intercept_)
```

63.976178067479765

```
In [24]: predict=en.predict(x_test)
```

```
In [25]: print(en.score(x_test,y_test))
```

0.40022118234614756

```
In [26]: from sklearn import metrics
```

```
In [27]: print("Mean Absolute error:",metrics.mean_absolute_error(y_test,predict))
```

Mean Absolute error: 43.914578294509376

```
In [28]: print("Mean Squared error:",metrics.mean_squared_error(y_test,predict))
```

Mean Squared error: 2994.8423056261063

```
In [29]: print("Root squared error:", np.sqrt(metrics.mean_squared_error(y_test, predict)))
```

Root squared error: 54.725152403863675

Model saving

```
In [30]: import pickle
filename="prediction"
pickle.dump(lr, open(filename, 'wb'))
filename='prediction'
model=pickle.load(open(filename, 'rb'))
```

```
In [31]: real=[[10],[7]]
result=model.predict(real)
result
```

Out[31]: array([63.98618302, 63.9831757])

In []: