DEENA 20104016

importing libraries

LINEAR REGRESSION

```
In [1]: import pandas as pd
import numpy as np
```

In [2]: data = pd.read_csv("20_states.csv")

Out[2]:

	id	name	country_id	country_code	country_name	state_code	type	latitude
0	3901	Badakhshan	1	AF	Afghanistan	BDS	NaN	36.734772
1	3871	Badghis	1	AF	Afghanistan	BDG	NaN	35.167134
2	3875	Baghlan	1	AF	Afghanistan	BGL	NaN	36.178903
3	3884	Balkh	1	AF	Afghanistan	BAL	NaN	36.755060
4	3872	Bamyan	1	AF	Afghanistan	BAM	NaN	34.810007
5072	1953	Mashonaland West Province	247	ZW	Zimbabwe	MW	NaN	-17.485103
5073	1960	Masvingo Province	247	ZW	Zimbabwe	MV	NaN	-20.624151
5074	1954	Matabeleland North Province	247	ZW	Zimbabwe	MN	NaN	-18.533157
5075	1952	Matabeleland South Province	247	ZW	Zimbabwe	MS	NaN	-21.052337
5076	1957	Midlands Province	247	ZW	Zimbabwe	MI	NaN	-19.055201

5077 rows × 9 columns

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Out[3]:

		id	name	country_id	country_code	country_name	state_code	type	latitude	lon
-	0	3901	Badakhshan	1	AF	Afghanistan	BDS	NaN	36.734772	70.8
	1	3871	Badghis	1	AF	Afghanistan	BDG	NaN	35.167134	63.7
	2	3875	Baghlan	1	AF	Afghanistan	BGL	NaN	36.178903	68.7
	3	3884	Balkh	1	AF	Afghanistan	BAL	NaN	36.755060	3.66
	4	3872	Bamyan	1	AF	Afghanistan	BAM	NaN	34.810007	67.8

In [4]:

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<pre>try_id country_code country_name \</pre>								
0	3901	Badakhshan	1	AF	Afghanistan			
1	3871	Badghis	1	AF	Afghanistan			
2	3875	Baghlan	1	AF	Afghanistan			
3	3884	Balkh	1	AF	Afghanistan			
4	3872	Bamyan	1	AF	Afghanistan			
		•••	• • •	• • •	• • •			
5072	1953	Mashonaland West Province	247	ZW	Zimbabwe			
5073	1960	Masvingo Province	247	ZW	Zimbabwe			
5074	1954	Matabeleland North Province	247	ZW	Zimbabwe			
5075	1952	Matabeleland South Province	247	ZW	Zimbabwe			
5076	1957	Midlands Province	247	ZW	Zimbabwe			

state_code	type	latitude	longitude
BDS	NaN	36.734772	70.811995
BDG	NaN	35.167134	63.769538
BGL	NaN	36.178903	68.745306
BAL	NaN	36.755060	66.897537
BAM	NaN	34.810007	67.821210
MW	NaN	-17.485103	29.788925
MV	NaN	-20.624151	31.262637
MN	NaN	-18.533157	27.549585
MS	NaN	-21.052337	29.045993
MI	NaN	-19.055201	29.603549
	BDS BDG BGL BAM MW MV MN	BDS NaN BDG NaN BGL NaN BAL NaN BAM NaN MW NaN MV NaN MN NaN MS NaN	BDS NaN 36.734772 BDG NaN 35.167134 BGL NaN 36.178903 BAL NaN 36.755060 BAM NaN 34.810007 MW NaN -17.485103 MV NaN -20.624151 MN NaN -18.533157 MS NaN -21.052337

[5077 rows x 9 columns]>

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In [5]:

Out[5]:

	id	country_id	latitude	longitude
count	5077.000000	5077.000000	5008.000000	5008.000000
mean	2609.765413	133.467599	27.576415	17.178713
std	1503.376799	72.341160	22.208161	61.269334
min	1.000000	1.000000	-54.805400	-178.116500
25%	1324.000000	74.000000	11.399747	-3.943859
50%	2617.000000	132.000000	34.226432	17.501792
75%	3905.000000	201.000000	45.802822	41.919647
max	5220.000000	248.000000	77.874972	179.852222

Train the model

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LASSO AND RIDGE

ELASTICNET

```
In [16]: from sklearn.linear_model import ElasticNet
a=ElasticNet()
Out[16]: ElasticNet()
```

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PREDICTION

```
In [19]: import pickle
    fn="prediction"

In [20]: import pandas as pd
    import pickle
    fn="prediction"

In [21]: r=[[10],[20]]
    result=m.predict(r)

Out[21]: array([10., 20.])
```

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