

DEENA 20104016

importing libraries

LINEAR REGRESSION

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

```
In [2]: data = pd.read_csv("19_nuclear_explosions.csv")
```

Out[2]:

	WEAPON SOURCE COUNTRY	WEAPON DEPLOYMENT LOCATION	Data.Source	Location.Cordinates.Latitude	Location.Cordinates.Lo
0	USA	Alamogordo	DOE	32.54	
1	USA	Hiroshima	DOE	34.23	
2	USA	Nagasaki	DOE	32.45	
3	USA	Bikini	DOE	11.35	
4	USA	Bikini	DOE	11.35	
...	
2041	CHINA	Lop Nor	HFS	41.69	
2042	INDIA	Pokhran	HFS	27.07	
2043	INDIA	Pokhran	NRD	27.07	
2044	PAKIST	Chagai	HFS	28.90	
2045	PAKIST	Kharan	HFS	28.49	

2046 rows × 6 columns

In [3]:

Out[3]:

	WEAPON SOURCE COUNTRY	WEAPON DEPLOYMENT LOCATION	Data.Source	Location.Cordinates.Latitude	Location.Cordinates.Long
0	USA	Alamogordo	DOE	32.54	-10
1	USA	Hiroshima	DOE	34.23	14
2	USA	Nagasaki	DOE	32.45	14
3	USA	Bikini	DOE	11.35	16
4	USA	Bikini	DOE	11.35	16

In [4]:

```

Out[4]: <bound method DataFrame.info of WEAPON SOURCE COUNTRY WEAPON DEPLOYMENT
LOCATION Data.Source \
0 USA Alamogordo DOE
1 USA Hiroshima DOE
2 USA Nagasaki DOE
3 USA Bikini DOE
4 USA Bikini DOE
...
2041 CHINA Lop Nor HFS
2042 INDIA Pokhran HFS
2043 INDIA Pokhran NRD
2044 PAKIST Chagai HFS
2045 PAKIST Kharan HFS

Location.Cordinates.Latitude Location.Cordinates.Longitude \
0 32.54 -105.57
1 34.23 132.27
2 32.45 129.52
3 11.35 165.20
4 11.35 165.20
...
2041 41.69 88.35
2042 27.07 71.70
2043 27.07 71.70
2044 28.90 64.89
2045 28.49 63.78

Data.Magnitude.Body Data.Magnitude.Surface Location.Cordinates.Depth
\
0 0.0 0.0 -0.10
1 0.0 0.0 -0.60
2 0.0 0.0 -0.60
3 0.0 0.0 -0.20
4 0.0 0.0 0.03
...
2041 5.3 0.0 0.00
2042 5.3 0.0 0.00
2043 0.0 0.0 0.00
2044 0.0 0.0 0.00
2045 5.0 0.0 0.00

Data.Yeild.Lower Data.Yeild.Upper Data.Purpose Data.Name Data.Type
\
0 21.0 21.0 Wr Trinity Tower
1 15.0 15.0 Combat Littleboy Airdrop
2 21.0 21.0 Combat Fatman Airdrop
3 21.0 21.0 We Able Airdrop
4 21.0 21.0 We Baker Uw
...
2041 3.0 12.0 Wr Nan Ug
2042 0.0 20.0 Wr Shakti 1-3 Ug
2043 0.0 1.0 Wr Nan Ug
2044 0.0 35.0 Wr Nan Ug
2045 0.0 18.0 Wr Nan Ug

```

	Date.Day	Date.Month	Date.Year
0	16	7	1945
1	5	8	1945
2	9	8	1945
3	30	6	1946
4	24	7	1946
...
2041	29	7	1996
2042	11	5	1998
2043	13	5	1998
2044	28	5	1998
2045	30	5	1998

[2046 rows x 16 columns]>

In [5]:

Out[5]:

	Location.Cordinates.Latitude	Location.Cordinates.Longitude	Data.Magnitude.Body	Data.Ma
count	2046.000000	2046.000000	2046.000000	
mean	35.462429	-36.015037	2.145406	
std	23.352702	100.829355	2.625453	
min	-49.500000	-169.320000	0.000000	
25%	37.000000	-116.051500	0.000000	
50%	37.100000	-116.000000	0.000000	
75%	49.870000	78.000000	5.100000	
max	75.100000	179.220000	7.400000	

Train the model

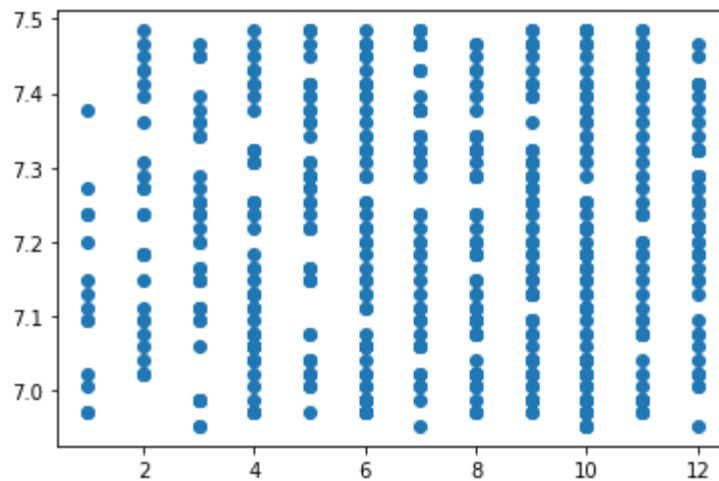
In [6]: `x = data[['Date.Day']]`In [7]: `# to split my dataset into training and test data`
`from sklearn.model_selection import train_test_split`In [8]: `from sklearn.linear_model import LinearRegression`
`lr = LinearRegression()`Out[8]: `LinearRegression()`In [9]: `coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])`

Out[9]:

	Co-efficient
Date.Day	-0.017747

```
In [10]: prediction= lr.predict(x_test)
```

```
Out[10]: <matplotlib.collections.PathCollection at 0x277a610d910>
```



```
In [11]:
```

```
Out[11]: -0.008014992423696743
```

LASSO AND RIDGE

```
In [12]: from sklearn.linear_model import Ridge,Lasso  
rr=Ridge(alpha=10)
```

```
Out[12]: Ridge(alpha=10)
```

```
In [13]:
```

```
Out[13]: -0.00801464851449385
```

```
In [14]: la=Lasso(alpha=10)
```

```
Out[14]: Lasso(alpha=10)
```

```
In [15]:
```

```
Out[15]: -0.006835547905959194
```

ELASTICNET

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

```
Out[16]: ElasticNet()
```

```
In [17]: print(a.coef_)
print(a.intercept_)
print(a.score(x_test,y_test))
```

```
[-0.01120072]
7.393213180123642
-0.006968863432721628
[7.12439579 7.09079362 7.38201246 7.23640304 7.37081173 7.23640304
 7.37081173 7.10199435 7.09079362 7.15799797 7.16919869 7.09079362
 7.32600883 7.13559652 7.23640304 7.09079362 7.20280086 7.04599072
 7.19160014 7.11319507 7.33720956 7.20280086 7.38201246 7.22520231
 7.12439579 7.18039942 7.15799797 7.0795929 7.25880449 7.11319507
 7.14679724 7.05719145 7.10199435 7.09079362 7.14679724 7.23640304
 7.22520231 7.23640304 7.12439579 7.22520231 7.23640304 7.27000521
 7.28120594 7.13559652 7.20280086 7.16919869 7.06839217 7.05719145
 7.22520231 7.34841028 7.19160014 7.31480811 7.21400159 7.28120594
 7.18039942 7.18039942 7.22520231 7.10199435 7.32600883 7.27000521
 7.27000521 7.15799797 7.04599072 7.15799797 7.35961101 7.16919869
 7.12439579 7.13559652 7.10199435 7.15799797 7.09079362 7.23640304
 7.22520231 7.29240666 7.24760376 7.28120594 7.06839217 7.20280086
 7.37081173 7.33720956 7.12439579 7.11319507 7.14679724 7.06839217
 7.35961101 7.21400159 7.30360738 7.12439579 7.19160014 7.0795929
 7.18039942 7.16919869 7.29240666 7.16919869 7.21400159 7.20280086
 7.21400159 7.14679724 7.11319507 7.24760376 7.20280086 7.22520231 7.22520231]
```

```
In [18]: from sklearn import metrics
print(" Mean Absolute Error :",metrics.mean_absolute_error(y_test,prediction))
print(" Mean Squared Error :",metrics.mean_squared_error(y_test,prediction))
```

```
Mean Absolute Error : 2.649414935840175
Mean Squared Error : 9.407836897692961
Root Mean Absolute Error : 1.6277023486621178
```

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([7.3247476 , 7.14727929])
```