## Vaibhav Kumar

**Roll No: 19** 

## Linear Regression, Weather Dataset, Weather Prediction Model

```
import pandas as pd
In [1]:
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        %matplotlib inline
        import seaborn as sns
        from sklearn.linear_model import LinearRegression
        from sklearn.metrics import mean_squared_error,mean_absolute_error
        from sklearn.model selection import train test split
        dfw=pd.read_csv('D:\\vk\\TRIM 3\\ML\\DATASET\\Weather.csv')
In [2]:
        C:\Users\studentadmin\AppData\Local\Temp\ipykernel_2960\2889413408.py:1: DtypeWarnin
        g: Columns (7,8,18,25) have mixed types. Specify dtype option on import or set low_me
        mory=False.
          dfw=pd.read_csv('D:\\vk\\TRIM 3\\ML\\DATASET\\Weather.csv')
In [3]:
        dfw
```

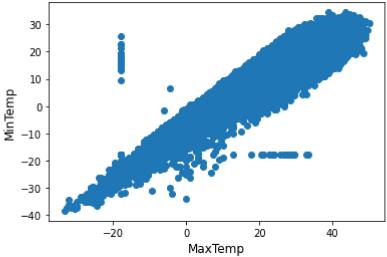
Out[3]:		STA	Date	Precip	WindGustSp	d Max	Temp	MinTem	o Mean	Temp	Snowfall	PoorWe	atŀ
	0	10001	1942- 7-1	1.016	Nai	N 25.5!	55556	22.22222	2 23.88	88889	0.0		N.
	1	10001	1942- 7-2	0	Nai	N 28.88	38889	21.66666	7 25.55	55556	0.0		N
	2	10001	1942- 7-3	2.54	Nai	N 26.1	11111	22.22222	2 24.44	44444	0.0		N.
	3	10001	1942- 7-4	2.54	Nai	V 26.66	66667	22.22222	2 24.44	44444	0.0		N
	4	10001	1942- 7-5	0	Nai	V 26.66	66667	21.66666	7 24.44	44444	0.0		N.
	•••												
	119035	82506	1945- 12-27	0	Nal	N 28.33	33333	18.33333	3 23.33	33333	0.0		N.
	119036	82506	1945- 12-28	9.906	Nai	N 29.44	14444	18.33333	3 23.88	88889	0.0		
	119037	82506	1945- 12-29	0	Nai	N 28.33	33333	18.33333	3 23.33	33333	0.0		
	119038	82506	1945- 12-30	0	Nal	N 28.33	33333	18.33333	3 23.33	33333	0.0		N
	119039	82506	1945- 12-31	0	Nal	N 29.44	14444	17.22222	2 23.33	33333	0.0		N.
	119040	rows × 3	31 colur	nns									
4													•
In [4]:	dfw.he	ad()											
Out[4]:	ST	A Date	Precip	) Wind	GustSpd Ma	хТетр	Min1	emp Me	anTemp	Snowf	fall Poor	Neather	Y
	<b>0</b> 1000	1942- 7-1		5	NaN 25.	555556	22.22	2222 23	3.888889	(	0.0	NaN	4
	<b>1</b> 1000	1942- 7-2		0	NaN 28.	888889	21.66	6667 25	5.555556	(	0.0	NaN	4
	<b>2</b> 1000	1942- 7-3		4	NaN 26.	111111	22.22	2222 24	1.444444	(	0.0	NaN	4
	<b>3</b> 1000	1942- 7-4	, , ,	4	NaN 26.	666667	22.22	2222 24	1.444444	(	0.0	NaN	4
	<b>4</b> 1000	1942- 7-5		)	NaN 26.	666667	21.66	6667 24	1.444444	(	0.0	NaN	4
	5 rows >	31 colu	umns										
4													•
In [5]:	dfw.de	scribe(	)										

	STA	WindGustSpd	MaxTemp	MinTemp	MeanTemp	YR	
count	119040.000000	532.000000	119040.000000	119040.000000	119040.000000	119040.000000	11
mean	29659.435795	37.774534	27.045111	17.789511	22.411631	43.805284	
std	20953.209402	10.297808	8.717817	8.334572	8.297982	1.136718	
min	10001.000000	18.520000	-33.333333	-38.333333	-35.55556	40.000000	
25%	11801.000000	29.632000	25.55556	15.000000	20.55556	43.000000	
50%	22508.000000	37.040000	29.444444	21.111111	25.555556	44.000000	
75%	33501.000000	43.059000	31.666667	23.333333	27.222222	45.000000	
max	82506.000000	75.932000	50.000000	34.444444	40.000000	45.000000	

8 rows × 24 columns

Out[5]:

```
In [6]: dfw.shape
Out[6]: (119040, 31)
In [7]: x=dfw['MaxTemp']
y=dfw['MinTemp']
In [8]: plt.scatter(x,y)
plt.xlabel('MaxTemp',fontsize='12')
plt.ylabel('MinTemp',fontsize='12')
plt.show()
```

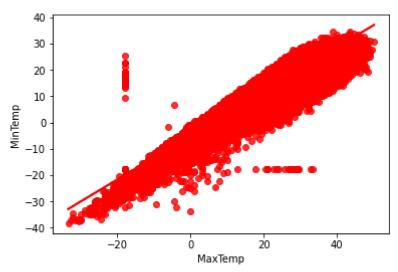


```
In [9]: sns.regplot(x,y,color='red')
```

D:\anaconda\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the foll owing variables as keyword args: x, y. From version 0.12, the only valid positional a rgument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

Out[9]: <AxesSubplot:xlabel='MaxTemp', ylabel='MinTemp'>

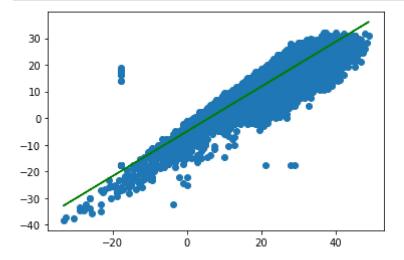


```
x.head()
In [10]:
               25.55556
Out[10]:
               28.888889
         2
               26.111111
         3
               26.666667
               26.666667
         Name: MaxTemp, dtype: float64
In [11]:
         y.head()
               22.22222
Out[11]:
               21.666667
          2
               22.22222
         3
               22.22222
               21.666667
         Name: MinTemp, dtype: float64
In [12]:
          x.shape
          (119040,)
Out[12]:
          X_=x.values.reshape(-1,1)
In [13]:
In [14]:
          X_.shape
          (119040, 1)
Out[14]:
In [15]:
                    25.55556
Out[15]:
                    28.888889
                    26.111111
          2
         3
                    26.666667
         4
                    26.666667
         119035
                    28.333333
         119036
                    29.44444
         119037
                    28.333333
         119038
                    28.333333
         119039
                    29.444444
         Name: MaxTemp, Length: 119040, dtype: float64
```

## Model

```
In [17]:
          X_train,X_test,y_train,y_test=train_test_split(X_,y,test_size=0.2,random_state=30)
In [18]:
          X_train.shape
          (95232, 1)
Out[18]:
In [19]:
          X_test.shape
          (23808, 1)
Out[19]:
In [20]:
          LR=LinearRegression()
          LR.fit(X_train,y_train)
         LinearRegression()
Out[20]:
         y_pred=LR.predict(X_test)
In [21]:
In [22]:
         y_test
         39071
                    12.22222
Out[22]:
         5109
                    22.22222
                    22.22222
         1113
         117003
                     7.777778
         106549
                    26.111111
                      . . .
         83309
                    12.22222
         75290
                     6.111111
         62785
                     8.888889
         80737
                     6.111111
         25569
                    22.777778
         Name: MinTemp, Length: 23808, dtype: float64
In [23]:
         y_pred
         array([14.66774639, 22.13310546, 18.86701087, ..., 15.13433134,
Out[23]:
                  9.53531203, 20.2667657 ])
In [24]:
          weights = LR.coef_
          intercept = LR.intercept_
          print(weights,intercept)
         [0.8398529] -4.928821159366091
In [25]:
          plt.scatter(X_test, y_test)
          plt.plot(X_test,y_pred, color='green')
```

```
plt.show()
```

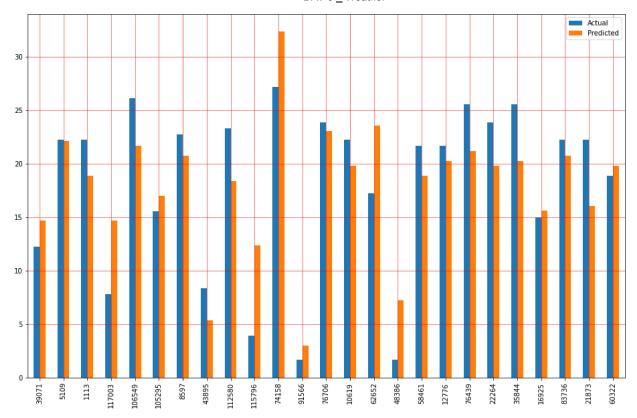


In [26]: df=pd.DataFrame({'Actual':y\_test,'Predicted':y\_pred})
 df

Out[26]:		Actual	Predicted
	39071	12.22222	14.667746
	5109	22.22222	22.133105
	1113	22.22222	18.867011
	117003	7.777778	14.667746
	106549	26.111111	21.666521
	•••		•••
	83309	12.22222	11.868237
	75290	6.111111	11.401652
	62785	8.888889	15.134331
	80737	6.111111	9.535312
	25569	22.777778	20.266766

23808 rows × 2 columns

```
In [27]: df1=df.head(25)
    df1.plot(kind='bar',figsize=(16,10))
    plt.grid(which='major',linestyle='-',linewidth='0.5',color='red')
    plt.grid(which='minor',linestyle=':',linewidth='0.5',color='green')
    plt.show()
```



```
In [28]: print('Mean Absolute Error', mean_absolute_error(y_test,y_pred))
    print('Mean Squared Error', mean_squared_error(y_test,y_pred))
    print('Root Mean Sqaured Error', np.sqrt(mean_squared_error(y_test,y_pred)))
```

Mean Absolute Error 3.086071147707306 Mean Squared Error 15.642509497194942 Root Mean Squured Error 3.955061250751364

**80:20** ::: Mean Absolute Error 3.086071147707306 Mean Squared Error 15.642509497194942 Root Mean Squared Error 3.955061250751364

**90:10** :: Mean Absolute Error 3.1047256786542192 Mean Squared Error 15.83161054325677 Root Mean Squared Error 3.97889564367511

**70:30** Mean Absolute Error 3.1013916713826886 Mean Squared Error 15.768060731561533 Root Mean Squared Error 3.970901752947501

## 80:20 split has most least error compared to others

In [ ]: