

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: df = pd.read_csv('temperatures.csv')
```

```
In [4]: df.head()
```

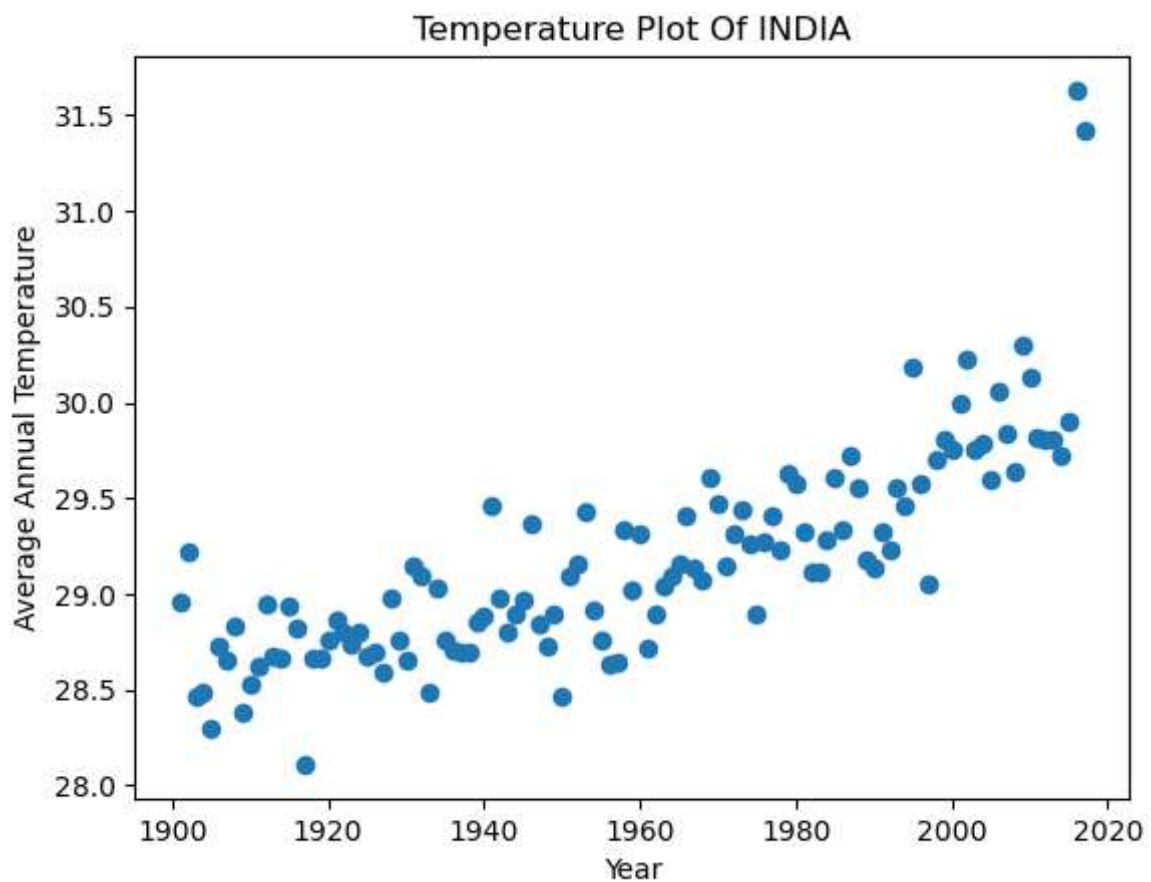
```
Out[4]:
```

	YEAR	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	ANNUAL
0	1901	22.40	24.14	29.07	31.91	33.41	33.18	31.21	30.39	30.47	29.97	27.31	24.49	28.96
1	1902	24.93	26.58	29.77	31.78	33.73	32.91	30.92	30.73	29.80	29.12	26.31	24.04	29.22
2	1903	23.44	25.03	27.83	31.39	32.91	33.00	31.34	29.98	29.85	29.04	26.08	23.65	28.47
3	1904	22.50	24.73	28.21	32.02	32.64	32.07	30.36	30.09	30.04	29.20	26.36	23.63	28.49
4	1905	22.00	22.83	26.68	30.01	33.32	33.25	31.44	30.68	30.12	30.67	27.52	23.82	28.30

```
In [5]: # input data
x = df['YEAR']
# output data
y = df['ANNUAL']
```

```
In [7]: #plt.figure(figsize=(16,9))
plt.title('Temperature Plot Of INDIA')
plt.xlabel('Year')
plt.ylabel('Average Annual Temperature')
plt.scatter(x, y)
```

```
Out[7]: <matplotlib.collections.PathCollection at 0x1a93ef9b090>
```



```
In [8]: x.shape
```

```
Out[8]: (117,)
```

```
In [9]: x = x.values
```

```
In [10]: x = x.reshape(117,1)
```

```
In [11]: x.shape
```

```
Out[11]: (117, 1)
```

```
In [12]: from sklearn.linear_model import LinearRegression
```

```
In [13]: regressor = LinearRegression()
```

```
In [14]: regressor.fit(x, y)
```

```
Out[14]: 

▼ LinearRegression



LinearRegression()


```

```
In [15]: regressor.coef_
```

```
Out[15]: array([0.01312158])
```

```
In [16]: regressor.intercept_
```

```
Out[16]: 3.4761897126187087
```

```
In [17]: regressor.predict([[2024]])
```

```
Out[17]: array([30.03427031])
```

```
In [18]: predicted = regressor.predict(x)
```

```
In [19]: predicted
```

```
Out[19]: array([28.4203158 , 28.43343739, 28.44655897, 28.45968055, 28.47280213,
 28.48592371, 28.49904529, 28.51216687, 28.52528846, 28.53841004,
 28.55153162, 28.5646532 , 28.57777478, 28.59089636, 28.60401794,
 28.61713952, 28.63026111, 28.64338269, 28.65650427, 28.66962585,
 28.68274743, 28.69586901, 28.70899059, 28.72211218, 28.73523376,
 28.74835534, 28.76147692, 28.7745985 , 28.78772008, 28.80084166,
 28.81396324, 28.82708483, 28.84020641, 28.85332799, 28.86644957,
 28.87957115, 28.89269273, 28.90581431, 28.91893589, 28.93205748,
 28.94517906, 28.95830064, 28.97142222, 28.9845438 , 28.99766538,
 29.01078696, 29.02390855, 29.03703013, 29.05015171, 29.06327329,
 29.07639487, 29.08951645, 29.10263803, 29.11575961, 29.1288812 ,
 29.14200278, 29.15512436, 29.16824594, 29.18136752, 29.1944891 ,
 29.20761068, 29.22073227, 29.23385385, 29.24697543, 29.26009701,
 29.27321859, 29.28634017, 29.29946175, 29.31258333, 29.32570492,
 29.3388265 , 29.35194808, 29.36506966, 29.37819124, 29.39131282,
 29.4044344 , 29.41755599, 29.43067757, 29.44379915, 29.45692073,
 29.47004231, 29.48316389, 29.49628547, 29.50940705, 29.52252864,
 29.53565022, 29.5487718 , 29.56189338, 29.57501496, 29.58813654,
 29.60125812, 29.6143797 , 29.62750129, 29.64062287, 29.65374445,
 29.66686603, 29.67998761, 29.69310919, 29.70623077, 29.71935236,
 29.73247394, 29.74559552, 29.7587171 , 29.77183868, 29.78496026,
 29.79808184, 29.81120342, 29.82432501, 29.83744659, 29.85056817,
 29.86368975, 29.87681133, 29.88993291, 29.90305449, 29.91617608,
 29.92929766, 29.94241924])
```

```
In [20]: y
```

```
Out[20]: 0      28.96
1      29.22
2      28.47
3      28.49
4      28.30
...
112    29.81
113    29.72
114    29.90
115    31.63
116    31.42
Name: ANNUAL, Length: 117, dtype: float64
```

```
In [21]: x
```

```
Out[21]: array([[1901],
               [1902],
               [1903],
               [1904],
               [1905],
               [1906],
               [1907],
               [1908],
               [1909],
               [1910],
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[2008],  
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[2010],  
[2011],  
[2012],  
[2013],  
[2014],  
[2015],  
[2016],  
[2017]], dtype=int64)
```

```
In [23]: # mean absolute error
```

```
abs(y - predicted)
```

```
Out[23]: 0      0.539684
         1      0.786563
         2      0.023441
         3      0.030319
         4      0.172802
         ...
        112     0.079933
        113     0.183054
        114     0.016176
        115     1.700702
        116     1.477581
        Name: ANNUAL, Length: 117, dtype: float64
```

```
In [24]: import numpy as np
```

```
In [25]: #mean absolute error
        np.mean(abs(y - predicted))
```

```
Out[25]: 0.22535284978630418
```

```
In [27]: from sklearn.metrics import mean_absolute_error
        mean_absolute_error(y, predicted)
```

```
Out[27]: 0.22535284978630418
```

```
In [28]: # mean squared error
        np.mean((y - predicted) ** 2)
```

```
Out[28]: 0.10960795229110358
```

```
In [31]: from sklearn.metrics import mean_squared_error
        mean_squared_error(y, predicted)
```

```
Out[31]: 0.10960795229110358
```

```
In [32]: from sklearn.metrics import r2_score
        r2_score(y, predicted)
```

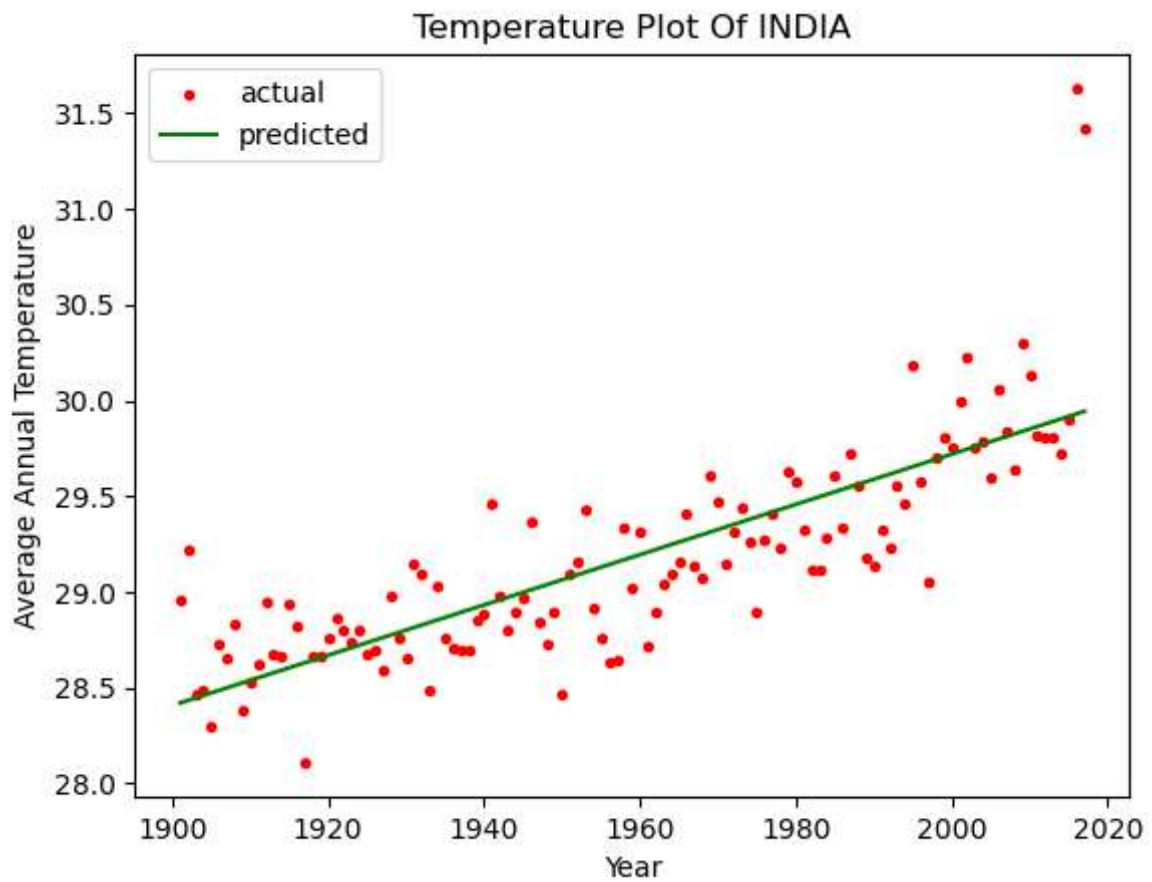
```
Out[32]: 0.641807891278368
```

```
In [34]: regressor.score(x, y)
```

```
Out[34]: 0.641807891278368
```

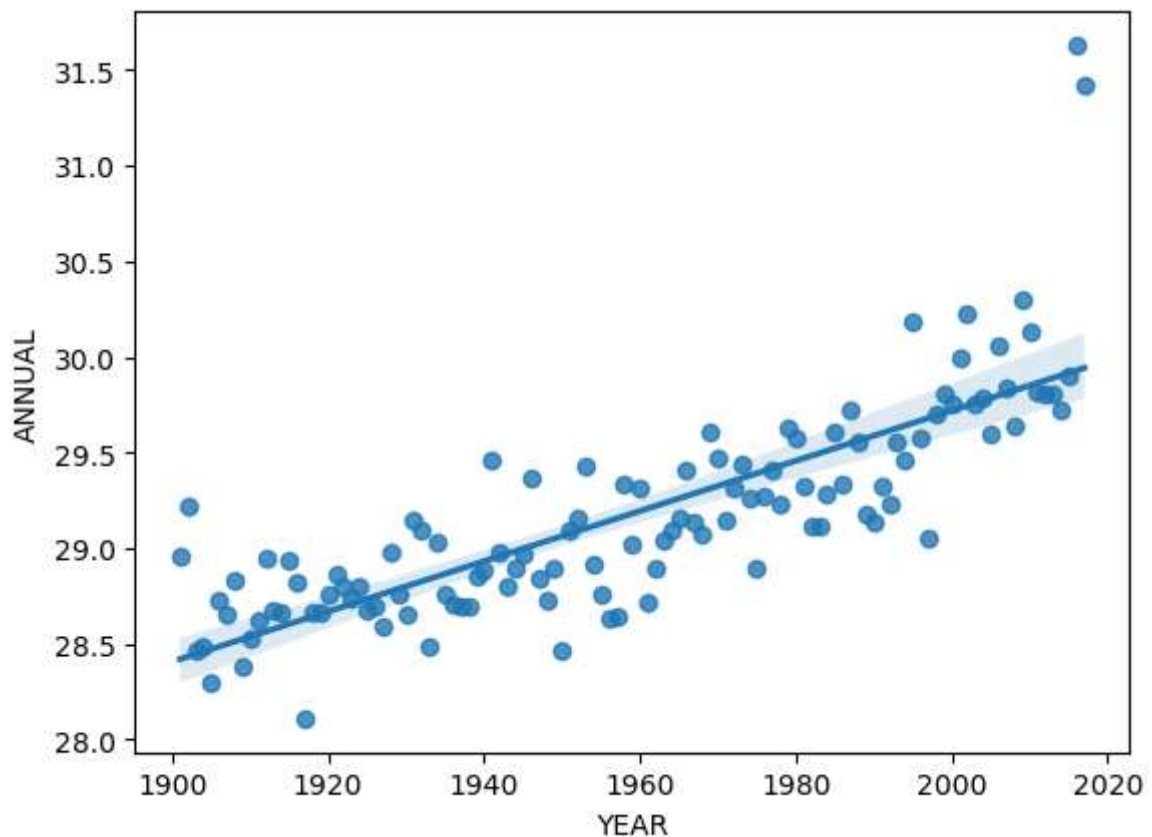
```
In [35]: #plt.figure(figsize=(16,9))
        plt.title('Temperature Plot Of INDIA')
        plt.xlabel('Year')
        plt.ylabel('Average Annual Temperature')
        plt.scatter(x, y, label = 'actual', color = 'r', marker = '.')
        plt.plot(x, predicted, label = 'predicted', color = 'g')
        plt.legend()
```

```
Out[35]: <matplotlib.legend.Legend at 0x1a93fb407d0>
```



```
In [36]: sns.regplot(x = 'YEAR', y = 'ANNUAL', data = df)
```

```
Out[36]: <Axes: xlabel='YEAR', ylabel='ANNUAL'>
```



In []: