```
#Name-Pawar ved balsaheb(T512037)
        import matplotlib.pyplot as plt
        import numpy as np
        import pandas as pd
        import seaborn as sns
        %matplotlib inline
        from sklearn.metrics import mean_squared_error,mean_absolute_error,r2_score
In [3]: temp_dataset = pd.read_csv('temperatures.csv')
        temp_dataset
Out[3]:
              YEAR
                     JAN
                            FEB MAR
                                        APR
                                              MAY
                                                     JUN
                                                            JUL
                                                                 AUG
                                                                        SEP
                                                                              OCT
                                                                                   NOV
              1901
                    22.40
                           24.14
                                 29.07
                                       31.91
                                                    33.18
                                                         31.21
                                                                 30.39
                                                                             29.97
                                              33.41
                                                                       30.47
                                                                                    27.31
                                                                                          2
                                                   32.91
              1902
                    24.93
                           26.58
                                 29.77
                                      31.78
                                             33.73
                                                          30.92
                                                                30.73
                                                                       29.80
                                                                             29.12
                                                                                    26.31
                                                                                         2
                    23.44
                          25.03 27.83 31.39
                                             32.91
                                                    33.00
                                                          31.34
                                                                 29.98
                                                                             29.04
                                                                                         2
              1903
                                                                       29.85
                                                                                    26.08
                                       32.02 32.64
                                                   32.07
              1904
                    22.50
                          24.73
                                 28.21
                                                          30.36
                                                                 30.09
                                                                       30.04
                                                                             29.20
                                                                                    26.36 2
              1905 22.00 22.83 26.68 30.01 33.32 33.25 31.44
                                                                30.68
                                                                      30.12 30.67
                                                                                    27.52 2
                                                                                          2
         112
              2013 24.56
                          26.59
                                30.62 32.66
                                             34.46
                                                   32.44
                                                          31.07 30.76
                                                                      31.04
                                                                             30.27
                                                                                    27.83
         113
                   23.83
                          25.97
                                 28.95 32.74 33.77 34.15 31.85
                                                                31.32 30.68
                                                                             30.29
                                                                                    28.05
                                                                                         2
              2014
         114
              2015 24.58
                          26.89
                                 29.07 31.87 34.09 32.48
                                                         31.88
                                                                31.52 31.55 31.04
                                                                                    28.10 2
         115
                          29.72 32.62 35.38 35.72 34.03 31.64
                                                                31.79 31.66
                                                                                    30.11 2
              2016 26.94
                                                                             31.98
              2017 26.45 29.46 31.60 34.95 35.84 33.82 31.88 31.72 32.22 32.29
         116
                                                                                    29.60 2
        117 rows × 18 columns
        temp_dataset.shape
In [5]:
Out[5]:
        (117, 18)
        temp_dataset.describe().T
In [7]:
```

Out[7]:

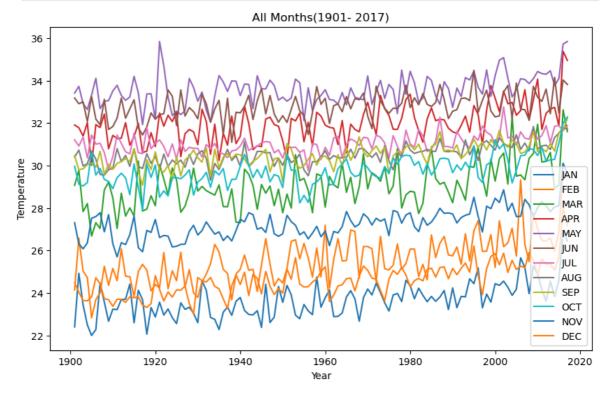
	count	mean	std	min	25%	50%	75%	max
YEAR	117.0	1959.000000	33.919021	1901.00	1930.00	1959.00	1988.00	2017.00
JAN	117.0	23.687436	0.834588	22.00	23.10	23.68	24.18	26.94
FEB	117.0	25.597863	1.150757	22.83	24.78	25.48	26.31	29.72
MAR	117.0	29.085983	1.068451	26.68	28.37	29.04	29.61	32.62
APR	117.0	31.975812	0.889478	30.01	31.46	31.95	32.42	35.38
MAY	117.0	33.565299	0.724905	31.93	33.11	33.51	34.03	35.84
JUN	117.0	32.774274	0.633132	31.10	32.34	32.73	33.18	34.48
JUL	117.0	31.035897	0.468818	29.76	30.74	31.00	31.33	32.76
AUG	117.0	30.507692	0.476312	29.31	30.18	30.54	30.76	31.84
SEP	117.0	30.486752	0.544295	29.07	30.12	30.52	30.81	32.22
ОСТ	117.0	29.766581	0.705492	27.90	29.38	29.78	30.17	32.29
NOV	117.0	27.285470	0.714518	25.70	26.79	27.30	27.72	30.11
DEC	117.0	24.608291	0.782644	23.02	24.04	24.66	25.11	28.01
ANNUAL	117.0	29.181368	0.55555	28.11	28.76	29.09	29.47	31.63
JAN-FEB	117.0	24.629573	0.911239	22.25	24.11	24.53	25.15	28.33
MAR- MAY	117.0	31.517607	0.740585	29.92	31.04	31.47	31.89	34.57
JUN-SEP	117.0	31.198205	0.420508	30.24	30.92	31.19	31.40	32.41
OCT- DEC	117.0	27.208120	0.672003	25.74	26.70	27.21	27.61	30.03

In [9]: temp_dataset.isnull().sum()

Out[9]: YEAR

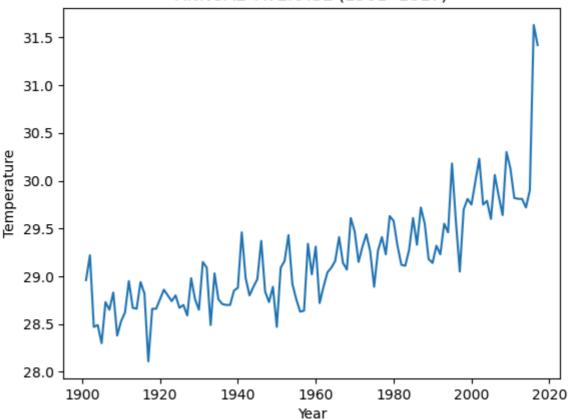
0 JAN 0 0 FEB MAR 0 0 APR 0 MAY JUN 0 JUL 0 AUG 0 SEP OCT 0 0 NOV DEC 0 0 ANNUAL JAN-FEB 0 $\mathsf{MAR}\text{-}\mathsf{MAY}$ JUN-SEP 0 OCT-DEC dtype: int64

```
In [11]: plt.figure(figsize=(10,6))
    plt.plot(temp_dataset["YEAR"],temp_dataset.iloc[:,1:13])
    plt.xlabel("Year")
    plt.ylabel("Temperature")
    plt.title("All Months(1901- 2017)")
    plt.legend(["JAN","FEB","MAR","APR","MAY","JUN","JUL","AUG","SEP","OCT","NOV","D
    plt.show()
```



```
In [13]: plt.plot(temp_dataset["YEAR"],temp_dataset["ANNUAL"])
    plt.xlabel("Year")
    plt.ylabel("Temperature")
    plt.title("ANNUAL AVERAGE (1901- 2017)")
    plt.show()
```

ANNUAL AVERAGE (1901-2017)



```
In [15]: from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression
         X=temp_dataset[["YEAR"]]
         y=temp_dataset["JAN"]
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
         print(X_train.shape, X_test.shape)
         # instantiate the model
         lr = LinearRegression()
         # fit the model
         lr.fit(X_train, y_train)
         #predicting the target value from the model for the samples
         y_test_lr = lr.predict(X_test)
         y_train_lr = lr.predict(X_train)
        (93, 1) (24, 1)
         print("Intercept", lr.intercept_)
In [17]:
         print("Slope", lr.coef_)
        Intercept 0.04088196282960865
        Slope [0.01206848]
In [19]: #computing the accuracy of the model performance
         acc_train_lr = lr.score(X_train, y_train)
         acc_test_lr = lr.score(X_test, y_test)
         #computing root mean squared error (RMSE)
         rmse_train_lr = np.sqrt(mean_squared_error(y_train, y_train_lr))
```

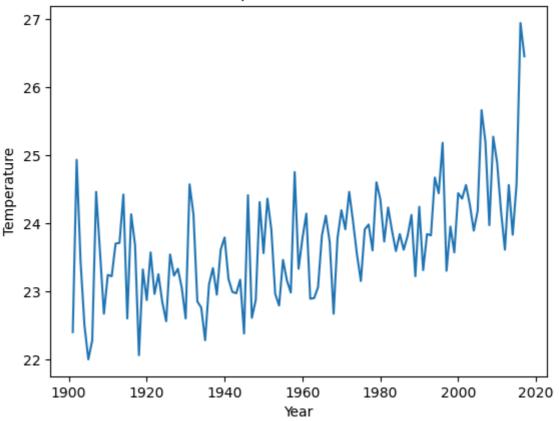
```
rmse_test_lr = np.sqrt(mean_squared_error(y_test, y_test_lr))
print("Linear Regression: Accuracy on training Data: {:.3f}".format(acc_train_lr)
print("Linear Regression: Accuracy on test Data: {:.3f}".format(acc_test_lr))
print('\nLinear Regression: The RMSE of the training set is:', rmse_train_lr)
print('Linear Regression: The RMSE of the testing set is:', rmse_test_lr)
```

Linear Regression: Accuracy on training Data: 0.264 Linear Regression: Accuracy on test Data: 0.460

Linear Regression: The RMSE of the training set is: 0.6660910902332355 Linear Regression: The RMSE of the testing set is: 0.7461610613861772

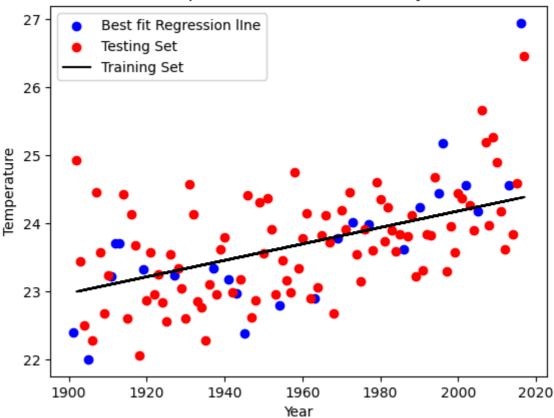
```
In [21]: #visulaize annul temperature
  plt.plot(X,y)
  plt.xlabel("Year")
  plt.ylabel("Temperature")
  plt.title("Annual Temperature from 1901-2017")
  plt.show()
```





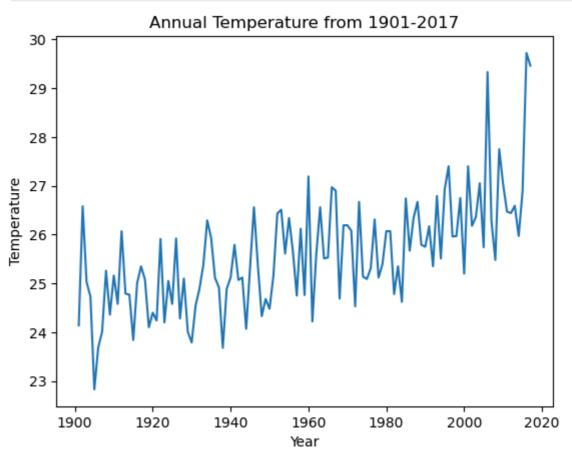
```
In [23]: # Visualization
#plt.figure(figsize=(8, 6))
plt.scatter(X_test,y_test,color = 'blue');
plt.scatter(X_train,y_train,color = 'red');
plt.plot(X_train,lr.predict(X_train), color = 'black');
plt.legend(['Best fit Regression lIne','Testing Set','Training Set'])
plt.title('Temperature vs Year for month Jan')
plt.xlabel('Year')
plt.ylabel('Temperature')
plt.show();
```

Temperature vs Year for month Jan



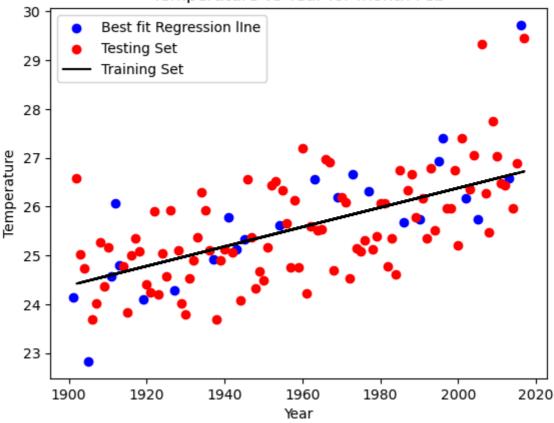
```
In [25]: #Errors for month Jan
         print('R-Squared Error :',r2_score(y_test,y_test_lr))
         print('Mean Absolute Error :',mean_absolute_error(y_test,y_test_lr))
         print('Mean Squared Error :',mean_squared_error(y_test,y_test_lr))
         print('Root Mean Squared Error :',np.sqrt(mean_squared_error(y_test,y_test_lr)))
        R-Squared Error: 0.4601171648515957
        Mean Absolute Error : 0.5083956669577453
        Mean Squared Error: 0.5567563295289465
        Root Mean Squared Error : 0.7461610613861772
In [27]: # Split the dataset into training and testing
         from sklearn.model selection import train test split
         from sklearn.linear_model import LinearRegression
         X=temp_dataset[["YEAR"]]
         y=temp_dataset["FEB"]
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
         print(X_train.shape, X_test.shape)
         # instantiate the model
         lr = LinearRegression()
         # fit the model
         lr.fit(X_train, y_train)
         #predicting the target value from the model for the samples
         y_test_lr = lr.predict(X_test)
         y_train_lr = lr.predict(X_train)
        (93, 1) (24, 1)
```

```
In [29]:
         print("Intercept", lr.intercept_)
         print("Slope", lr.coef_)
        Intercept -13.589454788386252
        Slope [0.01998518]
In [31]: #computing the accuracy of the model performance
         acc_train_lr = lr.score(X_train, y_train)
         acc_test_lr = lr.score(X_test, y_test)
         #computing root mean squared error (RMSE)
         rmse_train_lr = np.sqrt(mean_squared_error(y_train, y_train_lr))
         rmse_test_lr = np.sqrt(mean_squared_error(y_test, y_test_lr))
         print("Linear Regression: Accuracy on training Data: {:.3f}".format(acc_train_lr
         print("Linear Regression: Accuracy on test Data: {:.3f}".format(acc_test_lr))
         print('\nLinear Regression: The RMSE of the training set is:', rmse_train_lr)
         print('Linear Regression: The RMSE of the testing set is:', rmse_test_lr)
        Linear Regression: Accuracy on training Data: 0.366
        Linear Regression: Accuracy on test Data: 0.534
        Linear Regression: The RMSE of the training set is: 0.8686967555951215
        Linear Regression: The RMSE of the testing set is: 0.9079689831221699
         #visulaize annaul temperature
In [33]:
         plt.plot( X,y )
         plt.xlabel("Year")
         plt.ylabel("Temperature")
         plt.title("Annual Temperature from 1901-2017")
         plt.show()
```



```
In [35]: # Visualization
#plt.figure(figsize=(12, 9))
plt.scatter(X_test,y_test,color = 'blue');
plt.scatter(X_train,y_train,color = 'red');
plt.plot(X_train,lr.predict(X_train), color = 'black');
plt.legend(['Best fit Regression lIne','Testing Set','Training Set'])
plt.title('Temperature vs Year for month Feb')
plt.ylabel('Temperature')
plt.xlabel('Year')
plt.show();
```

Temperature vs Year for month Feb



```
In [37]: #Errors for month Feb
    print('R-Squared Error :',r2_score(y_test,y_test_lr))
    print('Mean Absolute Error :',mean_absolute_error(y_test,y_test_lr))
    print('Mean Squared Error :',mean_squared_error(y_test,y_test_lr))
    print('Root Mean Squared Error :',np.sqrt(mean_squared_error(y_test,y_test_lr)))

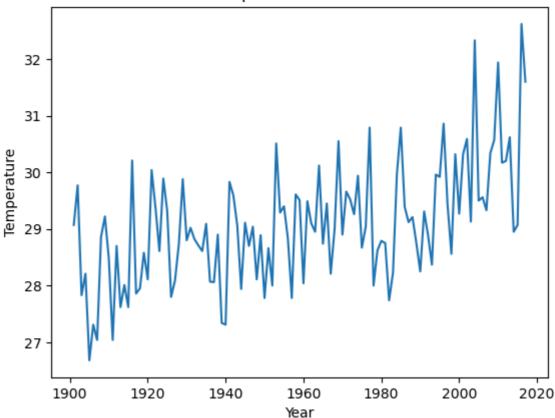
R-Squared Error : 0.5337774396814265
    Mean Absolute Error : 0.6321543551026766
    Mean Squared Error : 0.8244076743119072
    Root Mean Squared Error : 0.9079689831221699
```

```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression

X=temp_dataset[["YEAR"]]
y=temp_dataset["MAR"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando print(X_train.shape, X_test.shape)
```

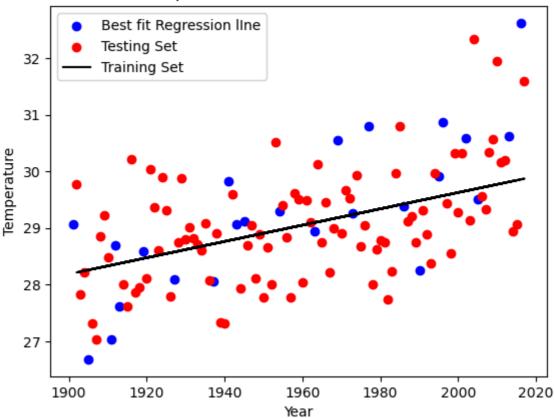
```
# instantiate the model
         lr = LinearRegression()
         # fit the model
         lr.fit(X_train, y_train)
         #predicting the target value from the model for the samples
         y_test_lr = lr.predict(X_test)
         y_train_lr = lr.predict(X_train)
        (93, 1) (24, 1)
In [41]: print("Intercept", lr.intercept_)
         print("Slope", lr.coef_)
        Intercept 0.8326151493664895
        Slope [0.01439656]
In [43]: #computing the accuracy of the model performance
         acc_train_lr = lr.score(X_train, y_train)
         acc_test_lr = lr.score(X_test, y_test)
         #computing root mean squared error (RMSE)
         rmse_train_lr = np.sqrt(mean_squared_error(y_train, y_train_lr))
         rmse_test_lr = np.sqrt(mean_squared_error(y_test, y_test_lr))
         print("Linear Regression: Accuracy on training Data: {:.3f}".format(acc train lr
         print("Linear Regression: Accuracy on test Data: {:.3f}".format(acc_test_lr))
         print('\nLinear Regression: The RMSE of the training set is:', rmse_train_lr)
         print('Linear Regression: The RMSE of the testing set is:', rmse_test_lr)
        Linear Regression: Accuracy on training Data: 0.234
        Linear Regression: Accuracy on test Data: 0.412
        Linear Regression: The RMSE of the training set is: 0.8613784589579908
        Linear Regression: The RMSE of the testing set is: 1.0069570422964147
In [45]: #visulaize annual temperature
         plt.plot( X,y )
         plt.xlabel("Year")
         plt.ylabel("Temperature")
         plt.title("Annual Temperature from 1901-2017")
         plt.show()
```

Annual Temperature from 1901-2017



```
In [47]: # Visualization
#plt.figure(figsize=(12, 9))
plt.scatter(X_test,y_test,color = 'blue');
plt.scatter(X_train,y_train,color = 'red');
plt.plot(X_train,lr.predict(X_train), color = 'black');
plt.legend(['Best fit Regression lIne','Testing Set','Training Set'])
plt.title('Temperature vs Year for month March')
plt.ylabel('Temperature')
plt.xlabel('Year')
plt.show();
```

Temperature vs Year for month March

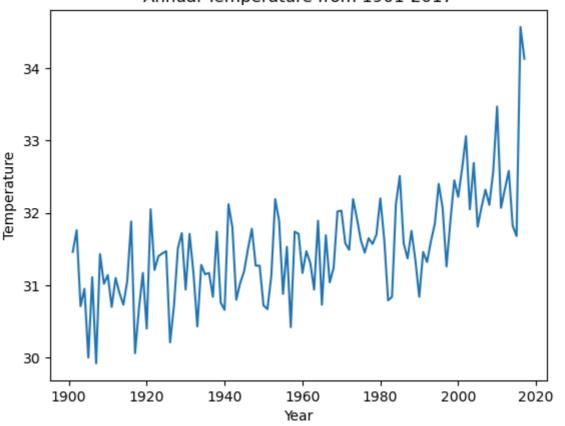


```
print('R-Squared Error :',r2_score(y_test,y_test_lr))
         print('Mean Absolute Error :',mean_absolute_error(y_test,y_test_lr))
         print('Mean Squared Error :',mean_squared_error(y_test,y_test_lr))
         print('Root Mean Squared Error :',np.sqrt(mean_squared_error(y_test,y_test_lr)))
        R-Squared Error: 0.41193423910621496
        Mean Absolute Error : 0.7784057105758042
        Mean Squared Error : 1.0139624850303435
        Root Mean Squared Error : 1.0069570422964147
In [51]: # Split the dataset into training and testing
         from sklearn.model selection import train test split
         from sklearn.linear_model import LinearRegression
         X=temp_dataset[["YEAR"]]
         y=temp dataset["MAR-MAY"]
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
         print(X_train.shape, X_test.shape)
         # instantiate the model
         lr = LinearRegression()
         # fit the model
         lr.fit(X_train, y_train)
         #predicting the target value from the model for the samples
         y_test_lr = lr.predict(X_test)
         y_train_lr = lr.predict(X_train)
        (93, 1) (24, 1)
```

In [49]: #Errors for month March

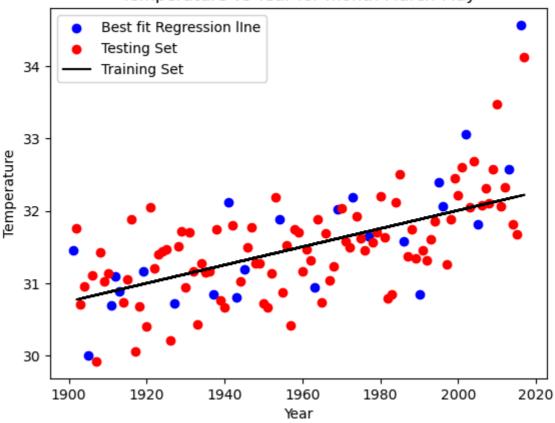
```
In [53]:
         print("Intercept", lr.intercept_)
         print("Slope", lr.coef_)
        Intercept 6.838794003180691
        Slope [0.01258417]
In [55]: #computing the accuracy of the model performance
         acc_train_lr = lr.score(X_train, y_train)
         acc_test_lr = lr.score(X_test, y_test)
         #computing root mean squared error (RMSE)
         rmse_train_lr = np.sqrt(mean_squared_error(y_train, y_train_lr))
         rmse_test_lr = np.sqrt(mean_squared_error(y_test, y_test_lr))
         print("Linear Regression: Accuracy on training Data: {:.3f}".format(acc_train_lr
         print("Linear Regression: Accuracy on test Data: {:.3f}".format(acc_test_lr))
         print('\nLinear Regression: The RMSE of the training set is:', rmse_train_lr)
         print('Linear Regression: The RMSE of the testing set is:', rmse_test_lr)
        Linear Regression: Accuracy on training Data: 0.381
        Linear Regression: Accuracy on test Data: 0.445
        Linear Regression: The RMSE of the training set is: 0.5297583992887077
        Linear Regression: The RMSE of the testing set is: 0.7002924995190984
In [57]: #visulaize annaul temperature
         plt.plot(X,y)
         plt.xlabel("Year")
         plt.ylabel("Temperature")
         plt.title("Annual Temperature from 1901-2017")
         plt.show()
```

Annual Temperature from 1901-2017



```
In [59]: # Visualization
#plt.figure(figsize=(12, 9))
plt.scatter(X_test,y_test,color = 'blue');
plt.scatter(X_train,y_train,color = 'red');
plt.plot(X_train,lr.predict(X_train), color = 'black');
plt.legend(['Best fit Regression lIne','Testing Set','Training Set'])
plt.title('Temperature vs Year for month March-May')
plt.ylabel('Temperature')
plt.xlabel('Year')
plt.show();
```

Temperature vs Year for month March-May



```
In [61]: #Errors for month Mar-May
    print('R-Squared Error :',r2_score(y_test,y_test_lr))
    print('Mean Absolute Error :',mean_absolute_error(y_test,y_test_lr))
    print('Mean Squared Error :',mean_squared_error(y_test,y_test_lr))
    print('Root Mean Squared Error :',np.sqrt(mean_squared_error(y_test,y_test_lr)))

R-Squared Error : 0.4448244662589811
    Mean Absolute Error : 0.5121496378249248
    Mean Squared Error : 0.49040958488270653
    Root Mean Squared Error : 0.7002924995190984
In []:
```