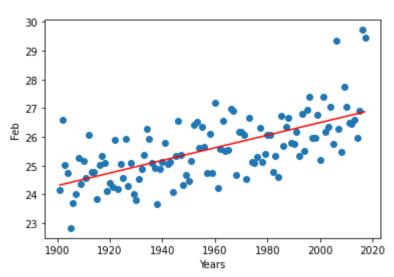
```
import pandas as pd
In [1]:
         import numpy as np
         df=pd.read csv(r'C:\Users\DELL\Desktop\temperatures.csv')
         df
              YEAR
                     JAN
                            FEB MAR
                                        APR MAY
                                                     JUN
                                                           JUL AUG
                                                                        SEP
                                                                              OCT NOV
                                                                                           DEC ANNUAL JAN-FEB MAR-MAY JUN-SEP OCT-DEC
Out[3]:
              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
                                                                                                             23.27
                                                                                                                       31.46
                                                                                                                                 31.27
                                                                                                                                           27.25
           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
                                                                                                             25.75
                                                                                                                       31.76
                                                                                                                                           26.49
                                                                                                                                 31.09
                                                                                                             24.24
                                                                                                                        30.71
                                                                                                                                           26.26
                     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
                                                                                                                                 30.92
               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
                                                                                                                        30.95
                                                                                                                                           26.40
           3
                                                                                                   28.49
                                                                                                             23.62
                                                                                                                                 30.66
                     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
                                                                                                             22.25
                                                                                                                        30.00
                                                                                                                                 31.33
                                                                                                                                           26.57
         112
               2013 24.56
                           26.59
                                 30.62 32.66 34.46
                                                   32.44 31.07 30.76 31.04
                                                                             30.27 27.83 25.37
                                                                                                   29.81
                                                                                                             25.58
                                                                                                                       32.58
                                                                                                                                 31.33
                                                                                                                                           27.83
               2014 23.83 25.97 28.95 32.74 33.77 34.15 31.85 31.32 30.68 30.29 28.05 25.08
                                                                                                   29.72
                                                                                                             24.90
                                                                                                                       31.82
                                                                                                                                 32.00
                                                                                                                                           27.81
         113
         114
               2015 24.58
                          26.89
                                 29.07
                                       31.87 34.09 32.48 31.88 31.52 31.55 31.04 28.10 25.67
                                                                                                   29.90
                                                                                                             25.74
                                                                                                                        31.68
                                                                                                                                 31.87
                                                                                                                                           28.27
         115
                    26.94
                           29.72 32.62
                                       35.38 35.72 34.03 31.64 31.79 31.66 31.98 30.11 28.01
                                                                                                   31.63
                                                                                                             28.33
                                                                                                                        34.57
                                                                                                                                 32.28
                                                                                                                                           30.03
         116
               2017 26.45 29.46 31.60 34.95 35.84 33.82 31.88 31.72 32.22 32.29 29.60 27.18
                                                                                                   31.42
                                                                                                             27.95
                                                                                                                        34.13
                                                                                                                                 32.41
                                                                                                                                           29.69
        117 rows × 18 columns
         from sklearn import linear model
         import matplotlib.pyplot as ply
In [5]:
         import matplotlib.pyplot as plt
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.JAN)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
```

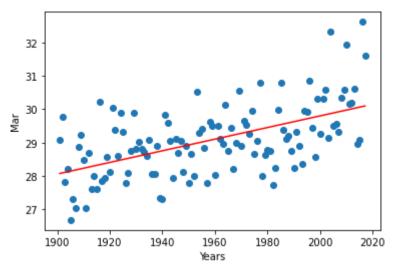
plt.xlabel("Years")

```
plt.ylabel("Jan")
         plt.scatter(df.YEAR,df.JAN)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
         C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
           warnings.warn(
        Predicted result = [24.50873533]
         [<matplotlib.lines.Line2D at 0x2319bedf670>]
Out[5]:
           27
           26
           25
         an
           24
           23
           22
                     1920
                             1940
                                            1980
                                                   2000
              1900
                                    1960
                                                           2020
                                    Years
        import matplotlib.pyplot as plt
In [6]:
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.FEB)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
         plt.xlabel("Years")
         plt.ylabel("Feb")
         plt.scatter(df.YEAR,df.FEB)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
        Predicted result = [26.8711227]
        C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
           warnings.warn(
```

```
Out[6]: [<matplotlib.lines.Line2D at 0x2319c39d610>]
```

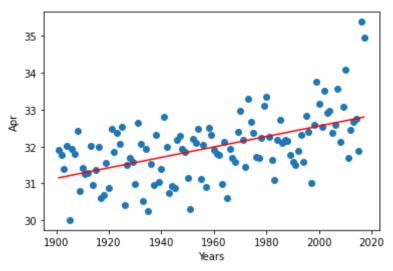


```
In [7]:
        import matplotlib.pyplot as plt
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.MAR)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
         plt.xlabel("Years")
         plt.ylabel("Mar")
         plt.scatter(df.YEAR,df.MAR)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
        Predicted result = [30.09793423]
        C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
          warnings.warn(
         [<matplotlib.lines.Line2D at 0x2319c41c3d0>]
Out[7]:
```

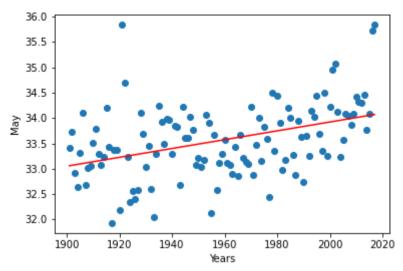


```
import matplotlib.pyplot as plt
In [8]:
        reg=linear model.LinearRegression()
        reg.fit(df[['YEAR']],df.APR)
        print("Predicted result =",reg.predict([[2017]]))
        %matplotlib inline
        plt.xlabel("Years")
        plt.ylabel("Apr")
        plt.scatter(df.YEAR,df.APR)
        plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
        Predicted result = [32.79814139]
        C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
        ssion was fitted with feature names
          warnings.warn(
        [<matplotlib.lines.Line2D at 0x2319c489130>]
```

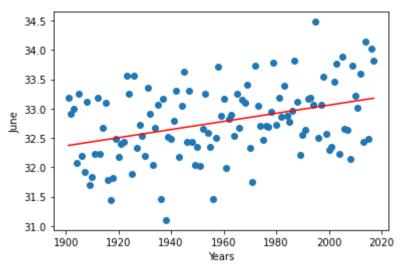
Out[8]:



Out[9]: [<matplotlib.lines.Line2D at 0x2319c4e5f10>]

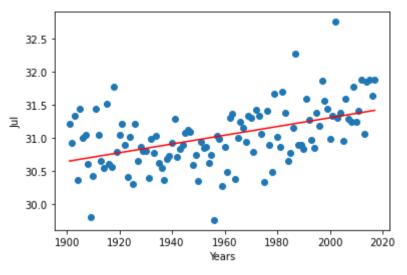


```
import matplotlib.pyplot as plt
In [10]:
          reg=linear model.LinearRegression()
          reg.fit(df[['YEAR']],df.JUN)
          print("Predicted result =",reg.predict([[2017]]))
          %matplotlib inline
          plt.xlabel("Years")
          plt.ylabel("June")
          plt.scatter(df.YEAR,df.JUN)
          plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
         Predicted result = [33.17683761]
         C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
          ssion was fitted with feature names
           warnings.warn(
          [<matplotlib.lines.Line2D at 0x2319c55afd0>]
Out[10]:
```

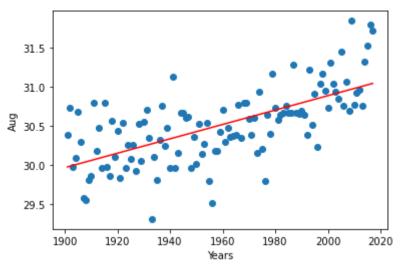


```
import matplotlib.pyplot as plt
In [11]:
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.JUL)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
         plt.xlabel("Years")
         plt.ylabel("Jul")
         plt.scatter(df.YEAR,df.JUL)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
         Predicted result = [31.41950022]
         C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
           warnings.warn(
         [<matplotlib.lines.Line2D at 0x2319c5d4400>]
```

Out[11]:



Out[12]: [<matplotlib.lines.Line2D at 0x2319c633d90>]

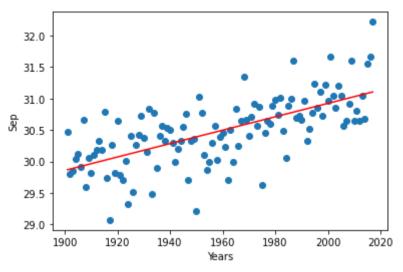


```
In [13]: import matplotlib.pyplot as plt
    reg=linear_model.LinearRegression()
    reg.fit(df[['YEAR']],df.SEP)
    print("Predicted result =",reg.predict([[2017]]))

%matplotlib inline
    plt.xlabel("Years")
    plt.ylabel("Sep")
    plt.ylabel("Sep")
    plt.scatter(df.YEAR,df.SEP)
    plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')

C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
    ssion was fitted with feature names
    warnings.warn(
    Predicted result = [31.10475735]

Out[13]: [<matplotlib.lines.Line2D at 0x2319c6a2af0>]
```



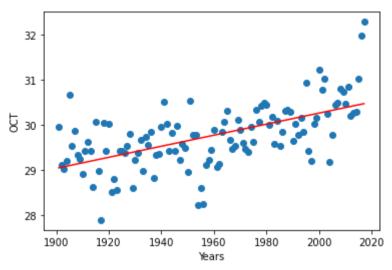
```
import matplotlib.pyplot as plt
reg=linear_model.LinearRegression()
reg.fit(df[['YEAR']],df.OCT)
print("Predicted result =",reg.predict([[2017]]))

/matplotlib inline
plt.xlabel("Years")
plt.ylabel("OCT")
plt.scatter(df.YEAR,df.OCT)
plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')

Predicted result = [30.4772157]

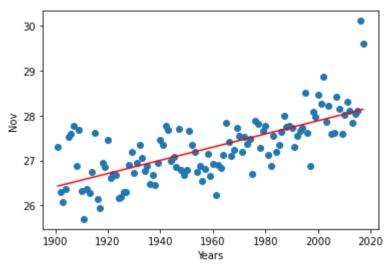
C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
ssion was fitted with feature names
warnings.warn(
```

Out[14]: [<matplotlib.lines.Line2D at 0x2319d6e4520>]

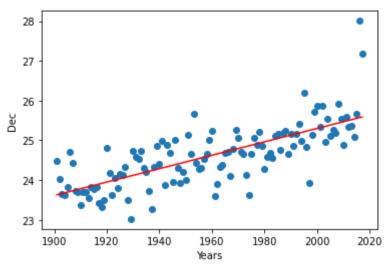


```
In [15]: import matplotlib.pyplot as plt
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.NOV)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
         plt.xlabel("Years")
         plt.ylabel("Nov")
         plt.scatter(df.YEAR,df.NOV)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
         Predicted result = [28.13814284]
         C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
           warnings.warn(
         [<matplotlib.lines.Line2D at 0x2319d73c730>]
```

Out[15]:

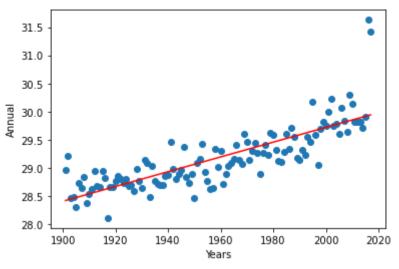


Out[16]: [<matplotlib.lines.Line2D at 0x2319d798910>]



```
In [17]: import matplotlib.pyplot as plt
         reg=linear model.LinearRegression()
         reg.fit(df[['YEAR']],df.ANNUAL)
         print("Predicted result =",reg.predict([[2017]]))
         %matplotlib inline
         plt.xlabel("Years")
         plt.ylabel("Annual")
         plt.scatter(df.YEAR,df.ANNUAL)
         plt.plot(df.YEAR,reg.predict(df[['YEAR']]),color='Red')
         Predicted result = [29.94241924]
         C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not have valid feature names, but LinearRegre
         ssion was fitted with feature names
           warnings.warn(
         [<matplotlib.lines.Line2D at 0x2319d805a30>]
```

Out[17]:



```
In [18]:
         import numpy as np
         from sklearn.metrics import mean absolute error, mean squared error
         from sklearn.model selection import train test split
         from sklearn.linear model import LinearRegression
         X=df[["YEAR"]]
         Y=df["JAN"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
         regressor = LinearRegression()
         regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
         rmse = np.sqrt(mse)
         print("Jan Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Jan Analysis
         Mean absolute error: 0.62
         Mean squared error: 0.61
         Root mean squared error: 0.78
In [19]: X=df[["YEAR"]]
         Y=df["FEB"]
         X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.3, random_state=0)
```

```
regressor = LinearRegression()
         regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
          rmse = np.sqrt(mse)
         print("Feb Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Feb Analysis
         Mean absolute error: 0.73
         Mean squared error: 0.95
         Root mean squared error: 0.97
In [20]: X=df[["YEAR"]]
         Y=df["MAR"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
          regressor = LinearRegression()
          regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared_error(Y_test, Y_pred)
         rmse = np.sqrt(mse)
         print("Mar Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Mar Analysis
         Mean absolute error: 0.70
         Mean squared error: 0.72
         Root mean squared error: 0.85
In [21]: X=df[["YEAR"]]
         Y=df["APR"]
         X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.3, random_state=0)
          regressor = LinearRegression()
         regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df_preds = pd.DataFrame({'Actual': Y_test.squeeze(), 'Predicted': Y_pred.squeeze()})
```

```
mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
         rmse = np.sqrt(mse)
         print("Apr Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Apr Analysis
         Mean absolute error: 0.58
         Mean squared error: 0.60
         Root mean squared error: 0.77
In [22]: X=df[["YEAR"]]
         Y=df["MAY"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
          regressor = LinearRegression()
          regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean_squared_error(Y_test, Y_pred)
         rmse = np.sqrt(mse)
         print("May Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         May Analysis
         Mean absolute error: 0.55
         Mean squared error: 0.47
         Root mean squared error: 0.68
In [23]: X=df[["YEAR"]]
         Y=df["JUN"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
         regressor = LinearRegression()
          regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df_preds = pd.DataFrame({'Actual': Y_test.squeeze(), 'Predicted': Y_pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
          rmse = np.sqrt(mse)
         print("Jun Analysis")
```

```
print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Jun Analysis
         Mean absolute error: 0.49
         Mean squared error: 0.35
         Root mean squared error: 0.59
In [24]: X=df[["YEAR"]]
         Y=df["JUL"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
         regressor = LinearRegression()
         regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y_test.squeeze(), 'Predicted': Y_pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
         rmse = np.sqrt(mse)
         print("July Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         July Analysis
         Mean absolute error: 0.37
         Mean squared error: 0.24
         Root mean squared error: 0.49
In [25]: X=df[["YEAR"]]
         Y=df["AUG"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
         regressor = LinearRegression()
          regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df_preds = pd.DataFrame({'Actual': Y_test.squeeze(), 'Predicted': Y_pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
         rmse = np.sqrt(mse)
         print("August Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
          print(f'Root mean squared error: {rmse:.2f}')
```

```
August Analysis
         Mean absolute error: 0.30
         Mean squared error: 0.13
         Root mean squared error: 0.36
In [26]: X=df[["YEAR"]]
         Y=df["SEP"]
         X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
          regressor = LinearRegression()
          regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
         rmse = np.sqrt(mse)
         print("Sept Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         Sept Analysis
         Mean absolute error: 0.33
         Mean squared error: 0.21
         Root mean squared error: 0.46
In [27]: X=df[["YEAR"]]
         Y=df["OCT"]
         X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.3, random_state=0)
          regressor = LinearRegression()
         regressor.fit(X train,Y train)
         Y pred = regressor.predict(X test)
         df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean squared error(Y test, Y pred)
          rmse = np.sqrt(mse)
         print("October Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
         October Analysis
         Mean absolute error: 0.40
         Mean squared error: 0.32
         Root mean squared error: 0.57
```

```
In [28]: X=df[["YEAR"]]
         Y=df["NOV"]
         X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.3, random_state=0)
          regressor = LinearRegression()
          regressor.fit(X train,Y train)
          Y pred = regressor.predict(X test)
          df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
          mae = mean absolute error(Y test, Y pred)
          mse = mean squared error(Y test, Y pred)
          rmse = np.sqrt(mse)
          print("November Analysis")
          print(f'Mean absolute error: {mae:.2f}')
          print(f'Mean squared error: {mse:.2f}')
          print(f'Root mean squared error: {rmse:.2f}')
         November Analysis
         Mean absolute error: 0.44
         Mean squared error: 0.32
          Root mean squared error: 0.57
In [29]: X=df[["YEAR"]]
         Y=df["DEC"]
          X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
          regressor = LinearRegression()
          regressor.fit(X train,Y train)
          Y pred = regressor.predict(X test)
          df preds = pd.DataFrame({'Actual': Y test.squeeze(), 'Predicted': Y pred.squeeze()})
          mae = mean absolute error(Y test, Y pred)
         mse = mean squared_error(Y_test, Y_pred)
          rmse = np.sqrt(mse)
          print("December Analysis")
          print(f'Mean absolute error: {mae:.2f}')
          print(f'Mean squared error: {mse:.2f}')
          print(f'Root mean squared error: {rmse:.2f}')
          December Analysis
         Mean absolute error: 0.35
         Mean squared error: 0.26
          Root mean squared error: 0.51
In [30]: X=df[["YEAR"]]
          Y=df["ANNUAL"]
          X train, X test, Y train, Y test = train test split(X,Y,test size=0.3, random state=0)
          regressor = LinearRegression()
```

```
regressor.fit(X train,Y train)
         Y_pred = regressor.predict(X_test)
         df preds = pd.DataFrame({'Actual': Y_test.squeeze(), 'Predicted': Y_pred.squeeze()})
         mae = mean absolute error(Y test, Y pred)
         mse = mean_squared_error(Y_test, Y_pred)
         rmse = np.sqrt(mse)
         print("Annual Analysis")
         print(f'Mean absolute error: {mae:.2f}')
         print(f'Mean squared error: {mse:.2f}')
         print(f'Root mean squared error: {rmse:.2f}')
        Annual Analysis
        Mean absolute error: 0.25
        Mean squared error: 0.15
        Root mean squared error: 0.38
In [ ]:
In [ ]:
In [ ]:
```