```
In [1]:
         import pandas as pd
         import numpy as np
         from sklearn.linear_model import LinearRegression
         import matplotlib.pyplot as plt
In [2]: # Load the data
         data = pd.read_csv('temperatures.csv')
In [3]: |data.isna().sum()
Out[3]: YEAR
                     0
         JAN
                     0
         FEB
                     0
         MAR
                     0
         APR
                     0
                     0
         MAY
         JUN
                     0
         JUL
                     0
         AUG
                     0
         SEP
                     0
                     0
         OCT
         NOV
                     0
         DEC
                     0
         ANNUAL
                     0
         JAN-FEB
                     0
         MAR-MAY
                     0
         JUN-SEP
                     0
         OCT-DEC
                     0
         dtype: int64
In [4]: data.head()
Out[4]:
            YEAR
                    JAN
                          FEB
                               MAR
                                      APR
                                                  JUN
                                                        JUL
                                                             AUG
                                                                    SEP
                                                                          OCT
                                                                                NOV
                                                                                      DEC ANN
                                            MAY
             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
                                                                                              2
          0
             1902 24.93 26.58 29.77 31.78
                                                 32.91
                                                       30.92
                                                             30.73 29.80 29.12 26.31
                                                                                               2
          1
                                          33.73
                                                                                     24.04
                         25.03 27.83 31.39
             1903 23.44
                                          32.91
                                                 33.00 31.34
                                                             29.98
                                                                   29.85
                                                                         29.04
                                                                               26.08 23.65
                                                                                               2
             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
                                                                                               2
             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
                                                                                               2
In [5]: |data.shape
Out[5]: (117, 18)
```

In [6]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 117 entries, 0 to 116
Data columns (total 18 columns):

| # | Column | Non-Null Count | Dtype | | | | | |
|-----------------------------|---------|----------------|---------|--|--|--|--|--|
| | | | | | | | | |
| 0 | YEAR | 117 non-null | int64 | | | | | |
| 1 | JAN | 117 non-null | float64 | | | | | |
| 2 | FEB | 117 non-null | float64 | | | | | |
| 3 | MAR | 117 non-null | float64 | | | | | |
| 4 | APR | 117 non-null | float64 | | | | | |
| 5 | MAY | 117 non-null | float64 | | | | | |
| 6 | JUN | 117 non-null | float64 | | | | | |
| 7 | JUL | 117 non-null | float64 | | | | | |
| 8 | AUG | 117 non-null | float64 | | | | | |
| 9 | SEP | 117 non-null | float64 | | | | | |
| 10 | ОСТ | 117 non-null | float64 | | | | | |
| 11 | NOV | 117 non-null | float64 | | | | | |
| 12 | DEC | 117 non-null | float64 | | | | | |
| 13 | ANNUAL | 117 non-null | float64 | | | | | |
| 14 | JAN-FEB | 117 non-null | float64 | | | | | |
| 15 | MAR-MAY | 117 non-null | float64 | | | | | |
| 16 | JUN-SEP | 117 non-null | float64 | | | | | |
| 17 | OCT-DEC | 117 non-null | float64 | | | | | |
| dtvnes float64(17) int64(1) | | | | | | | | |

dtypes: float64(17), int64(1)

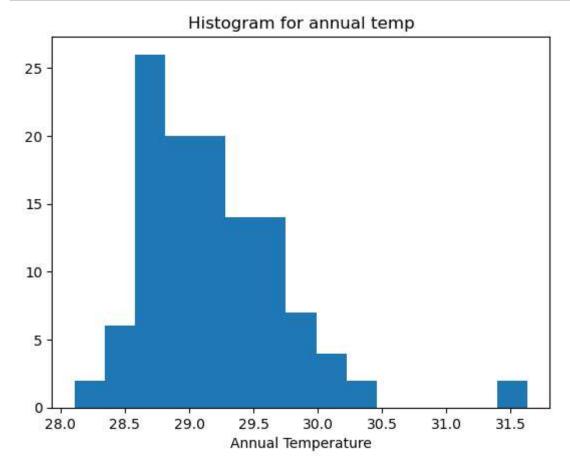
memory usage: 16.6 KB

In [7]: data.describe()

Out[7]:

| | YEAR | JAN | FEB | MAR | APR | MAY | JUN | |
|------|-------------|------------|------------|------------|------------|------------|------------|------|
| coun | 117.000000 | 117.000000 | 117.000000 | 117.000000 | 117.000000 | 117.000000 | 117.000000 | 117. |
| mear | 1959.000000 | 23.687436 | 25.597863 | 29.085983 | 31.975812 | 33.565299 | 32.774274 | 31. |
| sto | 33.919021 | 0.834588 | 1.150757 | 1.068451 | 0.889478 | 0.724905 | 0.633132 | 0. |
| mir | 1901.000000 | 22.000000 | 22.830000 | 26.680000 | 30.010000 | 31.930000 | 31.100000 | 29. |
| 25% | 1930.000000 | 23.100000 | 24.780000 | 28.370000 | 31.460000 | 33.110000 | 32.340000 | 30. |
| 50% | 1959.000000 | 23.680000 | 25.480000 | 29.040000 | 31.950000 | 33.510000 | 32.730000 | 31. |
| 75% | 1988.000000 | 24.180000 | 26.310000 | 29.610000 | 32.420000 | 34.030000 | 33.180000 | 31. |
| max | 2017.000000 | 26.940000 | 29.720000 | 32.620000 | 35.380000 | 35.840000 | 34.480000 | 32. |
| 4 | | | | | | | | • |

```
In [7]: data.shape
    plt.hist(data['ANNUAL'],bins = 15)
    plt.xlabel('Annual Temperature')
    plt.title('Histogram for annual temp')
    plt.show()
```

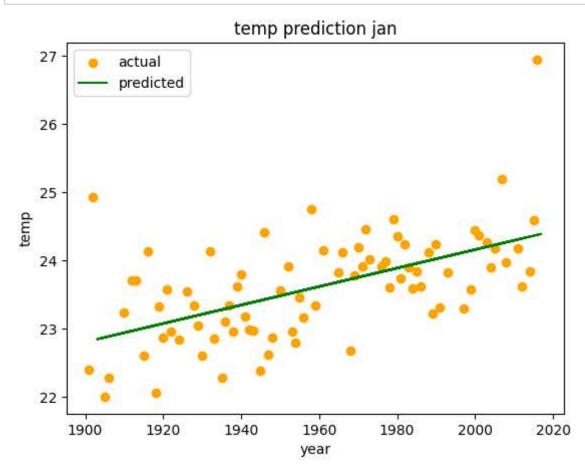


```
In [8]: X = data['YEAR'].values[:,None]
y = data['JAN']
```

```
In [19]: X
                 [1991],
                 [1992],
                 [1993],
                 [1994],
                 [1995],
                 [1996],
                 [1997],
                 [1998],
                 [1999],
                 [2000],
                 [2001],
                 [2002],
                 [2003],
                 [2004],
                 [2005],
                 [2006],
                 [2007],
                 [2008],
                 [2009],
                 [2010].
In [11]: y
Out[11]: 0
                 22.40
                 24.93
          1
          2
                 23.44
          3
                 22.50
          4
                 22.00
                 . . .
          112
                 24.56
          113
                 23.83
          114
                 24.58
                 26.94
          115
          116
                 26.45
          Name: JAN, Length: 117, dtype: float64
In [10]: |#Split the data into training and test datasets
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3,
          random state = 0)
```

```
In [11]: y_test
Out[11]: 10
                 23.22
          59
                 23.78
          95
                 25.18
          91
                 23.84
          7
                 23.57
          86
                 23.81
          26
                 23.23
          112
                 24.56
          22
                 23.25
          94
                 24.44
                 24.56
          101
          2
                 23.44
          50
                 24.36
          24
                 22.56
          116
                 26.45
          62
                 22.90
         93
                 24.67
          74
                 23.15
          61
                 22.89
          73
                 23.54
          16
                 23.68
          108
                 25.27
          13
                 24.42
          43
                 23.17
          105
                 25.66
          33
                 22.76
                 24.57
          30
          56
                 22.98
          48
                 24.31
          8
                 22.67
          97
                 23.95
                 23.72
          66
          109
                 24.89
          3
                 22.50
          63
                 23.06
                 24.46
          Name: JAN, dtype: float64
In [12]:
         # from sklearn.linear_model import train_test_split
          model1 = LinearRegression()
         model1.fit(X_train,y_train)
Out[12]:
          ▼ LinearRegression
          LinearRegression()
In [13]:
         y_test_predict = model1.predict(X_test)
          y_train_predict = model1.predict(X_train)
```

```
In [16]: plt.scatter(X_train,y_train,color = 'orange',label='actual')
    plt.plot(X_test,y_test_predict,color='green',label='predicted')
    plt.xlabel('year')
    plt.ylabel('temp')
    plt.title('temp prediction jan')
    plt.legend()
    plt.show()
```



```
In [14]:
         y_predict = model1.predict(X_test)
         y_predict
         y_test
Out[14]: 10
                 23.22
         59
                 23.78
         95
                 25.18
         91
                 23.84
         7
                 23.57
         86
                 23.81
                 23.23
         26
         112
                 24.56
         22
                 23.25
         94
                 24.44
         101
                 24.56
         2
                 23.44
                 24.36
         50
         24
                 22.56
         116
                 26.45
                 22.90
         62
         93
                 24.67
         74
                 23.15
         61
                 22.89
         73
                 23.54
         16
                 23.68
         108
                 25.27
         13
                 24.42
         43
                 23.17
         105
                 25.66
                 22.76
         33
         30
                 24.57
         56
                 22.98
         48
                 24.31
         8
                 22.67
         97
                 23.95
         66
                 23.72
         109
                 24.89
         3
                 22.50
                 23.06
         63
                 24.46
         Name: JAN, dtype: float64
In [15]: | from sklearn import metrics
         r_square = metrics.r2_score(y_test, y_test_predict)
         print('R-Square Error:', r_square)
```

R-Square Error: 0.27921723515312413

```
In [16]: from sklearn import metrics

mse = metrics.mean_squared_error(y_test, y_test_predict)
print('Mean Squared Error:', mse)

rmse = np.sqrt(mse)
print('Root Mean Squared Error:', rmse)

mae = metrics.mean_absolute_error(y_test, y_test_predict)
print('Mean Absolute Error:', mae)
```

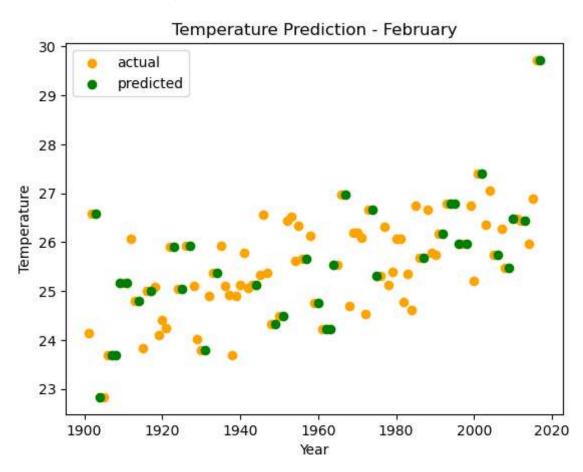
Mean Squared Error: 0.6080338203121165 Root Mean Squared Error: 0.7797652341006981 Mean Absolute Error: 0.6231302838065337

In []:

```
In [24]:
         import pandas as pd
         import numpy as np
         from sklearn.model_selection import train_test_split
         from sklearn.tree import DecisionTreeRegressor
         from sklearn import metrics
         import matplotlib.pyplot as plt
         # Load the data
         data = pd.read_csv('temperatures.csv')
         # Define a dictionary to map months to column names
         month_columns = {
              'JAN': 'January',
              'FEB': 'February',
              'MAR': 'March',
              'APR': 'April',
              'MAY': 'May',
              'JUN': 'June',
              'JUL': 'July',
              'AUG': 'August'
              'SEP': 'September',
              'OCT': 'October',
              'NOV': 'November'
              'DEC': 'December'
         }
         # Define a dictionary to map months to their corresponding number
         month_number_mapping = {
             1: 'JAN',
             2: 'FEB',
             3: 'MAR',
             4: 'APR',
             5: 'MAY',
             6: 'JUN',
             7: 'JUL',
             8: 'AUG',
             9: 'SEP',
             10: 'OCT',
             11: 'NOV',
             12: 'DEC'
         }
         # Get user input for month number
         while True:
             try:
                 month_number = int(input("Enter a month number (1-12): "))
                 if month number < 1 or month number > 12:
                      print("Please enter a valid month number between 1 and 12.")
                 else:
                      break
             except ValueError:
                  print("Invalid input. Please enter a valid number.")
         # Retrieve the corresponding month code and name using the dictionaries
         month_code = month_number_mapping[month_number]
         month_name = month_columns[month_code]
```

```
X = data['YEAR'].values.reshape(-1, 1)
y = data[month_code]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, rando
model = DecisionTreeRegressor()
model.fit(X_train, y_train)
y_test_predict = model.predict(X test)
y_train_predict = model.predict(X_train)
plt.scatter(X_train, y_train, color='orange', label='actual')
plt.scatter(X_test, y_test_predict, color='green', label='predicted')
plt.xlabel('Year')
plt.ylabel('Temperature')
plt.title(f'Temperature Prediction - {month_name}')
plt.legend()
plt.show()
# Calculate and print evaluation metrics
r_square = metrics.r2_score(y_test, y_test_predict)
mse = metrics.mean_squared_error(y_test, y_test_predict)
rmse = np.sqrt(mse)
mae = metrics.mean absolute error(y test, y test predict)
print(f'{month_name} - R-Square: {r_square:.2f}, MSE: {mse:.2f}, RMSE: {rmse:.
```

Enter a month number (1-12): 2



```
February - R-Square: 0.08, MSE: 1.66, RMSE: 1.29, MAE: 1.00
```

```
In [25]: from sklearn.tree import DecisionTreeRegressor
         # Create a Decision Tree Regressor model
         model = DecisionTreeRegressor()
         # Fit the model on the training data
         model.fit(X_train, y_train)
Out[25]:
          ▼ DecisionTreeRegressor
          DecisionTreeRegressor()
In [26]: predictions = model.predict(X_test)
In [28]: from sklearn.metrics import mean_squared_error, mean_absolute_error
         # ... (your code to train and predict using DecisionTreeRegressor) ...
         # Calculate evaluation metrics
         mse = mean_squared_error(y_test, predictions)
         rmse = np.sqrt(mse)
         mae = mean_absolute_error(y_test, predictions)
         print(f'MSE: {mse:.2f}, RMSE: {rmse:.2f}, MAE: {mae:.2f}')
         MSE: 1.66, RMSE: 1.29, MAE: 1.00
 In [ ]:
```