

```
import numpy as np;
import pandas as pd
import matplotlib.pyplot as plt;
import seaborn as sn

#load the dataset
dataset=pd.read_csv("c:\\Users\\lsrin\\Downloads\\TS-2\\Adavance_ML\\
climate.csv")
print(dataset)
```

	STATION	DATE	REPORT_TYPE	SOURCE
BackupElements \				
0	72518014735	2015-01-01T23:59:00	SOD	6
PRECIP				
1	72518014735	2015-01-02T23:59:00	SOD	6
PRECIP				
2	72518014735	2015-01-03T23:59:00	SOD	6
PRECIP				
3	72518014735	2015-01-04T23:59:00	SOD	6
PRECIP				
4	72518014735	2015-01-05T23:59:00	SOD	6
PRECIP				
...
...				
2663	72518014735	2022-05-27T23:59:00	SOD	6
PRECIP				
2664	72518014735	2022-05-28T23:59:00	SOD	6
PRECIP				
2665	72518014735	2022-05-29T23:59:00	SOD	6
PRECIP				
2666	72518014735	2022-05-30T23:59:00	SOD	6
PRECIP				
2667	72518014735	2022-05-31T23:59:00	SOD	6
PRECIP				

	BackupElevation	BackupEquipment	BackupLatitude	BackupLongitude
\				
0	260	PLASTIC	42.6918	-73.83109
1	260	PLASTIC	42.6918	-73.83109
2	260	PLASTIC	42.6918	-73.83109
3	260	PLASTIC	42.6918	-73.83109
4	260	PLASTIC	42.6918	-73.83109
...
2663	260	PLASTIC	42.6812	-73.81650

2664	260	PLASTIC	42.6812	-73.81650
2665	260	PLASTIC	42.6812	-73.81650
2666	260	PLASTIC	42.6812	-73.81650
2667	260	PLASTIC	42.6812	-73.81650
BackupName ... DailyPeakWindDirection DailyPeakWindSpeed				
\				
0	NWS ALBANY, NY	...	190.0	26.0
1	NWS ALBANY, NY	...	250.0	30.0
2	NWS ALBANY, NY	...	170.0	21.0
3	NWS ALBANY, NY	...	290.0	33.0
4	NWS ALBANY, NY	...	280.0	42.0
...
2663	NWS ALBANY, NY	...	160.0	28.0
2664	NWS ALBANY, NY	...	310.0	26.0
2665	NWS ALBANY, NY	...	90.0	13.0
2666	NWS ALBANY, NY	...	200.0	15.0
2667	NWS ALBANY, NY	...	250.0	29.0
DailyPrecipitation DailySnowDepth DailySnowfall \				
0	0.00	0.0	0.0	
1	T	0.0	T	
2	0.57	0.0	1.6	
3	0.22	1.0	0.0	
4	T	0.0	T	
...	
2663	0.00	0	0.0	
2664	0.04	0	0.0	
2665	0.00	0	0.0	
2666	0.00	0	0.0	
2667	0.00	0	0.0	
DailySustainedWindDirection DailySustainedWindSpeed Sunrise				
Sunset \				
0		190.0	20.0	726.0

1632.0			
1	310.0	23.0	726.0
1633.0			
2	160.0	15.0	726.0
1634.0			
3	290.0	24.0	726.0
1635.0			
4	290.0	32.0	726.0
1636.0			
...
...			
2663	160.0	21.0	423.0
1922.0			
2664	310.0	22.0	422.0
1923.0			
2665	180.0	9.0	421.0
1924.0			
2666	190.0	12.0	421.0
1925.0			
2667	250.0	21.0	420.0
1926.0			

	WindEquipmentChangeDate
0	2006-09-08
1	2006-09-08
2	2006-09-08
3	2006-09-08
4	2006-09-08
...	...
2663	2006-09-08
2664	2006-09-08
2665	2006-09-08
2666	2006-09-08
2667	2006-09-08

[2668 rows x 32 columns]

#DATA PROCESSING

dataset.head(10)

dataset.tail(8)

dataset.sample(7)

print(dataset.columns)

print(dataset.dtypes)

print(dataset.shape[0]) *#column*

print(dataset.shape[1]) *#rows*

print(dataset.size)

Index(['STATION', 'DATE', 'REPORT_TYPE', 'SOURCE', 'BackupElements',
'BackupElevation', 'BackupEquipment', 'BackupLatitude',
'BackupLongitude', 'BackupName',

```

'DailyAverageDewPointTemperature',
'DailyAverageDryBulbTemperature',
'DailyAverageRelativeHumidity',
'DailyAverageSeaLevelPressure', 'DailyAverageStationPressure',
'DailyAverageWetBulbTemperature', 'DailyAverageWindSpeed',
'DailyCoolingDegreeDays',
'DailyDepartureFromNormalAverageTemperature',
'DailyHeatingDegreeDays', 'DailyMaximumDryBulbTemperature',
'DailyMinimumDryBulbTemperature', 'DailyPeakWindDirection',
'DailyPeakWindSpeed', 'DailyPrecipitation', 'DailySnowDepth',
'DailySnowfall', 'DailySustainedWindDirection',
'DailySustainedWindSpeed', 'Sunrise', 'Sunset',
'WindEquipmentChangeDate'],
dtype='object')

```

```

STATION                int64
DATE                   object
REPORT_TYPE            object
SOURCE                 int64
BackupElements         object
BackupElevation        int64
BackupEquipment        object
BackupLatitude         float64
BackupLongitude        float64
BackupName             object
DailyAverageDewPointTemperature float64
DailyAverageDryBulbTemperature float64
DailyAverageRelativeHumidity float64
DailyAverageSeaLevelPressure float64
DailyAverageStationPressure float64
DailyAverageWetBulbTemperature float64
DailyAverageWindSpeed float64
DailyCoolingDegreeDays float64
DailyDepartureFromNormalAverageTemperature float64
DailyHeatingDegreeDays float64
DailyMaximumDryBulbTemperature float64
DailyMinimumDryBulbTemperature float64
DailyPeakWindDirection float64
DailyPeakWindSpeed float64
DailyPrecipitation     object
DailySnowDepth         object
DailySnowfall          object
DailySustainedWindDirection float64
DailySustainedWindSpeed float64
Sunrise                float64
Sunset                 float64
WindEquipmentChangeDate object
dtype: object
2668

```

```
32
85376
```

```
#do some statistical for int64,float64
dataNumerical=dataset.select_dtypes(include=['int64','float64'])
#print(dataNumerical.columns)
print(dataNumerical.dtypes)
```

STATION	int64
SOURCE	int64
BackupElevation	int64
BackupLatitude	float64
BackupLongitude	float64
DailyAverageDewPointTemperature	float64
DailyAverageDryBulbTemperature	float64
DailyAverageRelativeHumidity	float64
DailyAverageSeaLevelPressure	float64
DailyAverageStationPressure	float64
DailyAverageWetBulbTemperature	float64
DailyAverageWindSpeed	float64
DailyCoolingDegreeDays	float64
DailyDepartureFromNormalAverageTemperature	float64
DailyHeatingDegreeDays	float64
DailyMaximumDryBulbTemperature	float64
DailyMinimumDryBulbTemperature	float64
DailyPeakWindDirection	float64
DailyPeakWindSpeed	float64
DailySustainedWindDirection	float64
DailySustainedWindSpeed	float64
Sunrise	float64
Sunset	float64
dtype:	object

```
print(dataNumerical.describe())
```

	STATION	SOURCE	BackupElevation	BackupLatitude	
BackupLongitude \					
count	2.668000e+03	2668.0	2668.0	2668.000000	
2668.000000					
mean	7.251801e+10	6.0	260.0	42.689750	-
73.828268					
std	0.000000e+00	0.0	0.0	0.004187	
0.005764					
min	7.251801e+10	6.0	260.0	42.681200	-
73.831090					
25%	7.251801e+10	6.0	260.0	42.691800	-
73.831090					
50%	7.251801e+10	6.0	260.0	42.691800	-
73.831090					
75%	7.251801e+10	6.0	260.0	42.691800	-

```

73.831090
max      7.251801e+10      6.0      260.0      42.691800      -
73.816500

      DailyAverageDewPointTemperature      DailyAverageDryBulbTemperature
\
count      2668.000000      2668.000000
mean      38.217766      50.107571
std      19.116250      18.747310
min      -19.000000      -3.000000
25%      24.000000      35.000000
50%      38.000000      51.000000
75%      55.000000      67.000000
max      73.000000      87.000000

      DailyAverageRelativeHumidity      DailyAverageSeaLevelPressure      \
count      2668.000000      2668.000000
mean      66.085082      30.031945
std      13.401359      0.223771
min      24.000000      29.240000
25%      57.000000      29.880000
50%      66.000000      30.020000
75%      76.000000      30.180000
max      100.000000      30.740000

      DailyAverageStationPressure      ...      \
count      2668.000000      ...
mean      29.709059      ...
std      0.220846      ...
min      28.890000      ...
25%      29.570000      ...
50%      29.700000      ...
75%      29.850000      ...
max      30.420000      ...

      DailyDepartureFromNormalAverageTemperature
DailyHeatingDegreeDays      \
count      2668.000000
2668.000000
mean      2.155660
17.040480
std      8.202932

```

16.134205	
min	-28.700000
0.000000	
25%	-3.200000
0.000000	
50%	1.900000
14.000000	
75%	7.200000
30.000000	
max	34.700000
68.000000	

	DailyMaximumDryBulbTemperature	DailyMinimumDryBulbTemperature
\		
count	2668.000000	2668.000000
mean	59.418666	40.299100
std	20.003706	18.122395
min	5.000000	-13.000000
25%	42.000000	27.000000
50%	60.000000	40.000000
75%	77.000000	55.250000
max	97.000000	77.000000

	DailyPeakWindDirection	DailyPeakWindSpeed	\
count	2668.000000	2668.000000	
mean	222.387556	25.513493	
std	90.828564	9.436276	
min	10.000000	6.000000	
25%	170.000000	19.000000	
50%	260.000000	24.000000	
75%	290.000000	31.000000	
max	360.000000	70.000000	

	DailySustainedWindDirection	DailySustainedWindSpeed
Sunrise \		
count	2668.000000	2668.000000
2668.000000		
mean	223.924288	19.023238
563.145427		
std	90.846564	6.942113
108.536855		
min	10.000000	5.000000

416.000000		
25%	170.000000	14.000000
447.000000		
50%	270.000000	18.000000
547.000000		
75%	290.000000	23.000000
650.000000		
max	360.000000	67.000000
726.000000		

	Sunset
count	2668.000000
mean	1783.491004
std	111.230222
min	1621.000000
25%	1658.000000
50%	1805.000000
75%	1905.000000
max	1938.000000

[8 rows x 23 columns]

```
# # print(dataNumerical.info())
# dataNumerical.isnull().sum()
# print(dataNumerical.columns) #drop the
Station,source,'BackupElevation',
'BackupLatitude','BackupLongitude',Sunrise', 'Sunset

# dataFront=dataNumerical.iloc[:,0:5]
# dataBack=dataNumerical.iloc[:, -2:]
# print(dataFront.columns)
# print(dataBack.columns)
# # Combine the column names from dataFront and dataBack
# # columns_to_drop = list(dataFront.columns) + list(dataBack.columns)
# # #now drop those variable from NumericalDate
# # dataNumerical=dataNumerical.drop(columns=columns_to_drop)
# # print("The Final dataset to work:",dataNumerical.columns)
```

```
dataNumerical.isnull().sum()
```

STATION	0
SOURCE	0
BackupElevation	0
BackupLatitude	0
BackupLongitude	0
DailyAverageDewPointTemperature	0
DailyAverageDryBulbTemperature	0
DailyAverageRelativeHumidity	0
DailyAverageSeaLevelPressure	0
DailyAverageStationPressure	0


```

DailyAverageWetBulbTemperature      0
DailyAverageWindSpeed                0
DailyCoolingDegreeDays              0
DailyDepartureFromNormalAverageTemperature  0
DailyHeatingDegreeDays              0
DailyMaximumDryBulbTemperature      0
DailyMinimumDryBulbTemperature      0
DailyPeakWindDirection              0
DailyPeakWindSpeed                  0
DailySustainedWindDirection         0
DailySustainedWindSpeed              0
Sunrise                             0
Sunset                              0
dtype: int64

```

```
# Print the column names of the DataFrame
```

```
print(dataNumerical.columns)
```

```
# Drop the specified columns from the DataFrame
```

```

dataNumerical = dataNumerical.drop(columns=[
    'STATION', 'SOURCE', 'BackupElevation', 'BackupLatitude',
    'BackupLongitude', 'Sunrise', 'Sunset'
])

```

```

Index(['STATION', 'SOURCE', 'BackupElevation', 'BackupLatitude',
      'BackupLongitude', 'DailyAverageDewPointTemperature',
      'DailyAverageDryBulbTemperature',
      'DailyAverageRelativeHumidity',
      'DailyAverageSeaLevelPressure', 'DailyAverageStationPressure',
      'DailyAverageWetBulbTemperature', 'DailyAverageWindSpeed',
      'DailyCoolingDegreeDays',
      'DailyDepartureFromNormalAverageTemperature',
      'DailyHeatingDegreeDays', 'DailyMaximumDryBulbTemperature',
      'DailyMinimumDryBulbTemperature', 'DailyPeakWindDirection',
      'DailyPeakWindSpeed', 'DailySustainedWindDirection',
      'DailySustainedWindSpeed', 'Sunrise', 'Sunset'],
      dtype='object')

```

	DailyAverageDewPointTemperature	DailyAverageDryBulbTemperature
0	7.0	25.0
1	17.0	32.0
2	18.0	27.0
3	35.0	39.0
4	11.0	27.0
...

2663	63.0	70.0
2664	59.0	68.0
2665	50.0	66.0
2666	59.0	72.0
2667	62.0	79.0

	DailyAverageRelativeHumidity	DailyAverageSeaLevelPressure \
0	46.0	29.97
1	57.0	30.18
2	74.0	30.46
3	86.0	29.76
4	59.0	30.12
...
2663	76.0	29.83
2664	72.0	29.74
2665	57.0	30.03
2666	62.0	30.03
2667	55.0	29.93

	DailyAverageStationPressure	DailyAverageWetBulbTemperature \
0	29.65	21.0
1	29.80	26.0
2	30.16	23.0
3	29.47	38.0
4	29.73	20.0
...
2663	29.54	66.0
2664	29.41	63.0
2665	29.70	58.0
2666	29.73	64.0
2667	29.62	68.0

	DailyAverageWindSpeed	DailyCoolingDegreeDays \
0	8.8	0.0
1	9.5	0.0
2	4.3	0.0
3	10.0	0.0
4	16.8	0.0
...
2663	12.7	5.0
2664	7.6	3.0
2665	2.2	1.0
2666	3.4	7.0
2667	8.4	14.0

DailyDepartureFromNormalAverageTemperature
 DailyHeatingDegreeDays \

0	1.4
40.0	
1	8.6
33.0	
2	3.7
38.0	
3	15.9
26.0	
4	4.1
38.0	
...	...
...	
2663	8.8
0.0	
2664	6.5
0.0	
2665	4.2
0.0	
2666	9.9
0.0	
2667	16.5
0.0	

DailyMaximumDryBulbTemperature
 DailyMinimumDryBulbTemperature \

0	32.0	18.0
1	37.0	26.0
2	33.0	20.0
3	45.0	33.0
4	41.0	13.0
...
2663	77.0	63.0
2664	80.0	56.0
2665	80.0	51.0
2666	88.0	55.0
2667	92.0	65.0

	DailyPeakWindDirection	DailyPeakWindSpeed
DailySustainedWindDirection \		
0	190.0	26.0
190.0		
1	250.0	30.0
310.0		
2	170.0	21.0
160.0		
3	290.0	33.0
290.0		
4	280.0	42.0
290.0		
...
...		
2663	160.0	28.0
160.0		
2664	310.0	26.0
310.0		
2665	90.0	13.0
180.0		
2666	200.0	15.0
190.0		
2667	250.0	29.0
250.0		

	DailySustainedWindSpeed
0	20.0
1	23.0
2	15.0
3	24.0
4	32.0
...	...
2663	21.0
2664	22.0
2665	9.0
2666	12.0
2667	21.0

[2668 rows x 16 columns]

Display the updated DataFrame

`print(dataNumerical.columns)`

```
Index(['DailyAverageDewPointTemperature',
      'DailyAverageDryBulbTemperature',
      'DailyAverageRelativeHumidity', 'DailyAverageSeaLevelPressure',
      'DailyAverageStationPressure',
      'DailyAverageWetBulbTemperature',
      'DailyAverageWindSpeed', 'DailyCoolingDegreeDays',
      'DailyDepartureFromNormalAverageTemperature',
```

```
'DailyHeatingDegreeDays',
    'DailyMaximumDryBulbTemperature',
'DailyMinimumDryBulbTemperature',
    'DailyPeakWindDirection', 'DailyPeakWindSpeed',
    'DailySustainedWindDirection', 'DailySustainedWindSpeed'],
dtype='object')
```

using linearRegression

```
''' Predict Precipitation based on Temperature: - X (Feature):
DailyAverageDryBulbTemperature - Y (Target): DailyPrecipitation '''
```

using Multi-linear Regression

```
# X all except "DailyAverageDryBulbTemperature"
```

using multiple (ploynomial Regression)

```
# Using Linear Regress
# Features matrix (X) with the temperature column
X = dataNumerical[['DailyAverageDryBulbTemperature']]

# Ensure 'DailyPrecipitation' is numeric and fill NaN values with 0
dataNumerical['DailyPrecipitation'] =
pd.to_numeric(dataNumerical['DailyPrecipitation'], errors='coerce')
dataNumerical['DailyPrecipitation'].fillna(0, inplace=True)

# Target vector (y) with precipitation values
y = dataNumerical['DailyPrecipitation']

# Print X and y to verify
print(X)
print(y)
```

	DailyAverageDryBulbTemperature
0	25.0
1	32.0
2	27.0
3	39.0
4	27.0
...	...
2663	70.0
2664	68.0
2665	66.0

```
2666          72.0
2667          79.0
```

```
[2668 rows x 1 columns]
```

```
0      0.00
1      0.00
2      0.57
3      0.22
4      0.00
```

```
...
2663    0.00
2664    0.04
2665    0.00
2666    0.00
2667    0.00
```

```
Name: DailyPrecipitation, Length: 2668, dtype: float64
```

C:\Users\lsrin\AppData\Local\Temp\ipykernel_13048\1432735839.py:6:
FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
dataNumerical['DailyPrecipitation'].fillna(0, inplace=True)
```

```
from sklearn.model_selection import train_test_split
```

```
# Split the data into training and testing sets (80% train, 20% test)
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y,  
test_size=0.2, random_state=32)
```

```
# Print the number of rows in X_train
```

```
print(X_train.shape[0])
```

```
2134
```

```
#train the model
```

```
from sklearn.linear_model import LinearRegression
```

```
# Initialize and train the linear regression model
```

```
model = LinearRegression()
```

```
model.fit(X_train, y_train)
```

```
LinearRegression()
```

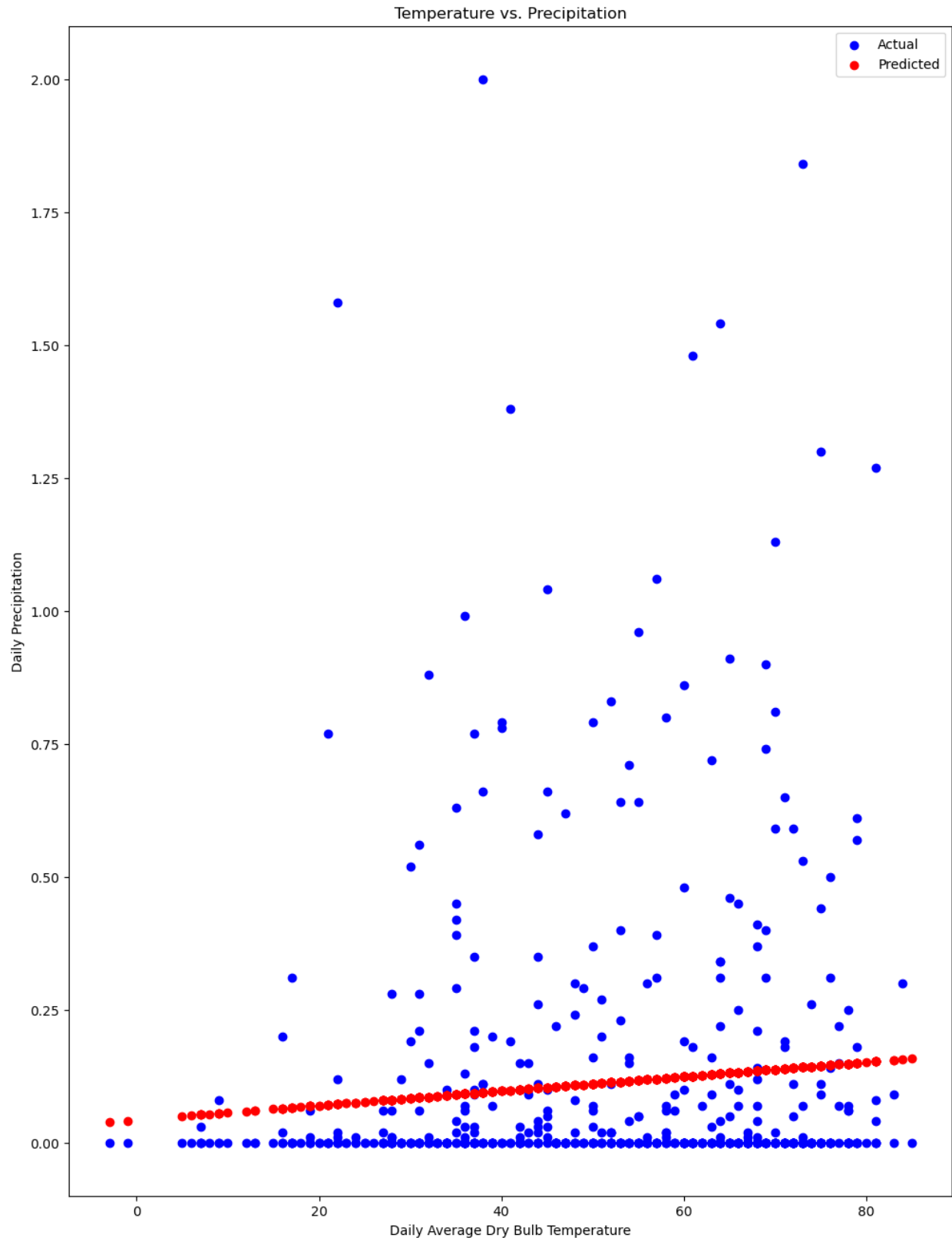
```
# Make predictions
y_pred = model.predict(X_test)
#y_pred

# Evaluate the model
from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)

print(f'Mean Squared Error: {mse}')
print(f'R-squared: {r2}')
```

Mean Squared Error: 0.0729113990218066
R-squared: 0.0025889789634933047

```
#Plot the true vs. predicted values.
import matplotlib.pyplot as plt
plt.figure(figsize=(12, 16))
plt.scatter(X_test, y_test, color='blue', label='Actual')
plt.scatter(X_test, y_pred, color='red', label='Predicted')
plt.xlabel('Daily Average Dry Bulb Temperature')
plt.ylabel('Daily Precipitation')
plt.title('Temperature vs. Precipitation')
plt.legend()
plt.show()
```



```
dataNumerical.columns
```



```
Index(['DailyAverageDewPointTemperature',
      'DailyAverageDryBulbTemperature',
      'DailyAverageRelativeHumidity', 'DailyAverageSeaLevelPressure',
      'DailyAverageStationPressure',
      'DailyAverageWetBulbTemperature',
      'DailyAverageWindSpeed', 'DailyCoolingDegreeDays',
      'DailyDepartureFromNormalAverageTemperature',
      'DailyHeatingDegreeDays',
      'DailyMaximumDryBulbTemperature',
      'DailyMinimumDryBulbTemperature',
      'DailyPeakWindDirection', 'DailyPeakWindSpeed',
      'DailySustainedWindDirection', 'DailySustainedWindSpeed',
      'DailyPrecipitation'],
      dtype='object')
```

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import PolynomialFeatures, StandardScaler
from sklearn.linear_model import Lasso, Ridge
from sklearn.metrics import mean_squared_error, r2_score
```

```
# Step 1: Data Preparation
```

```
X = data.drop(columns=['DailyPrecipitation']) # Features
y = data['DailyPrecipitation'] # Target
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=32)
```

```
# Step 2: Polynomial Transformation (Degree 2)
```

```
poly = PolynomialFeatures(degree=2, include_bias=False)
X_train_poly = poly.fit_transform(X_train)
X_test_poly = poly.transform(X_test)
```

```
# Step 3: Feature Scaling
```

```
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train_poly)
X_test_scaled = scaler.transform(X_test_poly)
```

```
# Lasso Regression with increased max_iter
```

```
lasso = Lasso(alpha=0.1, max_iter=10000)
lasso.fit(X_train_scaled, y_train)
y_pred_lasso = lasso.predict(X_test_scaled)
```

```
# Ridge Regression with increased max_iter
```

```
ridge = Ridge(alpha=0.1, max_iter=10000)
ridge.fit(X_train_scaled, y_train)
y_pred_ridge = ridge.predict(X_test_scaled)
```

```
# Step 4: Evaluate Lasso Model
```

```
mse_lasso = mean_squared_error(y_test, y_pred_lasso)
r2_lasso = r2_score(y_test, y_pred_lasso)
```

```

print(f'Lasso Regression - MSE: {mse_lasso}, R²: {r2_lasso}')

# Step 4: Evaluate Ridge Model
mse_ridge = mean_squared_error(y_test, y_pred_ridge)
r2_ridge = r2_score(y_test, y_pred_ridge)
print(f'Ridge Regression - MSE: {mse_ridge}, R²: {r2_ridge}')

Lasso Regression - MSE: 0.06687719302293675, R²: 0.08513551691536181
Ridge Regression - MSE: 0.048396740676427255, R²: 0.33794381700895426

#actual vs predication
import matplotlib.pyplot as plt
import numpy as np

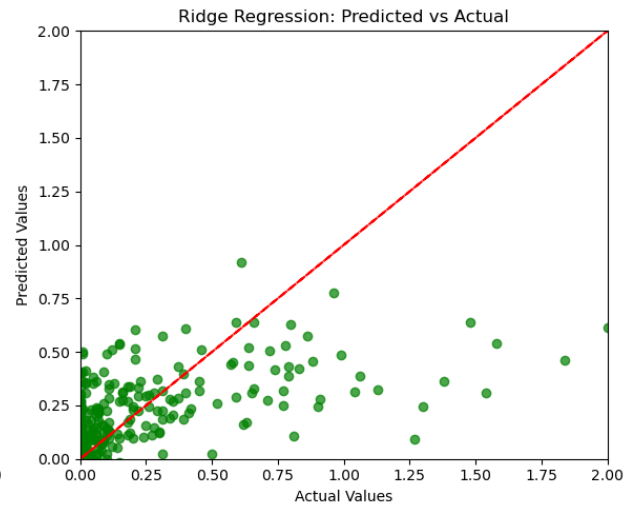
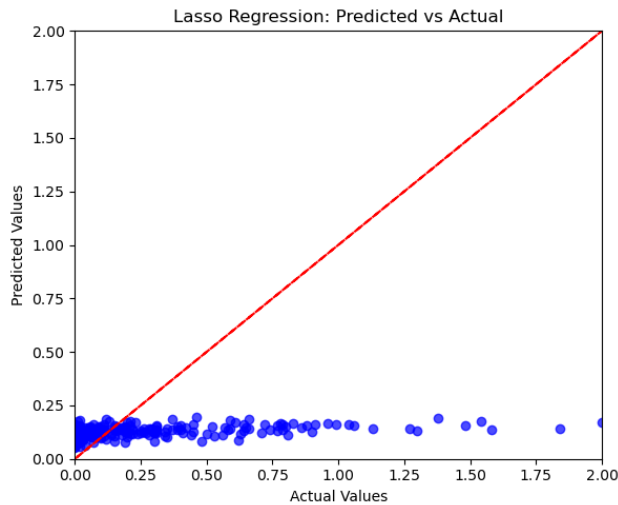
# Scatter plot of predicted vs actual for Lasso
plt.figure(figsize=(12, 5))

# Lasso Regression
plt.subplot(1, 2, 1)
plt.scatter(y_test, y_pred_lasso, color='blue', alpha=0.7)
plt.plot(y_test, y_test, color='red', linestyle='--') # Perfect
prediction line
plt.title('Lasso Regression: Predicted vs Actual')
plt.xlabel('Actual Values')
plt.ylabel('Predicted Values')
plt.xlim([0, max(y_test)])
plt.ylim([0, max(y_test)])

# Ridge Regression
plt.subplot(1, 2, 2)
plt.scatter(y_test, y_pred_ridge, color='green', alpha=0.7)
plt.plot(y_test, y_test, color='red', linestyle='--') # Perfect
prediction line
plt.title('Ridge Regression: Predicted vs Actual')
plt.xlabel('Actual Values')
plt.ylabel('Predicted Values')
plt.xlim([0, max(y_test)])
plt.ylim([0, max(y_test)])

plt.tight_layout()
plt.show()

```



```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression

# Use one feature for visualization (e.g.,
# DailyAverageDryBulbTemperature)
X_single_feature = data[['DailyAverageDryBulbTemperature']]
y = data['DailyPrecipitation']

# Split the data
X_train_single, X_test_single, y_train_single, y_test_single =
train_test_split(X_single_feature, y, test_size=0.2, random_state=32)

# Create polynomial features
poly = PolynomialFeatures(degree=2) # Change the degree as needed
X_train_poly = poly.fit_transform(X_train_single)
X_test_poly = poly.transform(X_test_single)

# Fit a Linear Regression model
model = LinearRegression()
model.fit(X_train_poly, y_train_single)

# Create a range of values for plotting the polynomial curve
X_plot = np.linspace(X_single_feature.min(), X_single_feature.max(),
100).reshape(-1, 1)
X_plot_poly = poly.transform(X_plot)
y_plot = model.predict(X_plot_poly)

# Plot the actual data and the polynomial regression curve
plt.figure(figsize=(10, 6))
plt.scatter(X_single_feature, y, color='blue', label='Actual Data',
alpha=0.6)
plt.plot(X_plot, y_plot, color='red', label='Polynomial Regression
```

```
Curve', linewidth=2)
plt.title('Polynomial Regression Visualization')
plt.xlabel('Daily Average Dry Bulb Temperature')
plt.ylabel('Daily Precipitation')
# plt.legend()
# plt.grid()
# plt.show()
```

C:\Users\lsrin\anaconda3\Lib\site-packages\sklearn\base.py:493:
 UserWarning: X does not have valid feature names, but
 PolynomialFeatures was fitted with feature names
 warnings.warn(

```
Text(0, 0.5, 'Daily Precipitation')
```

