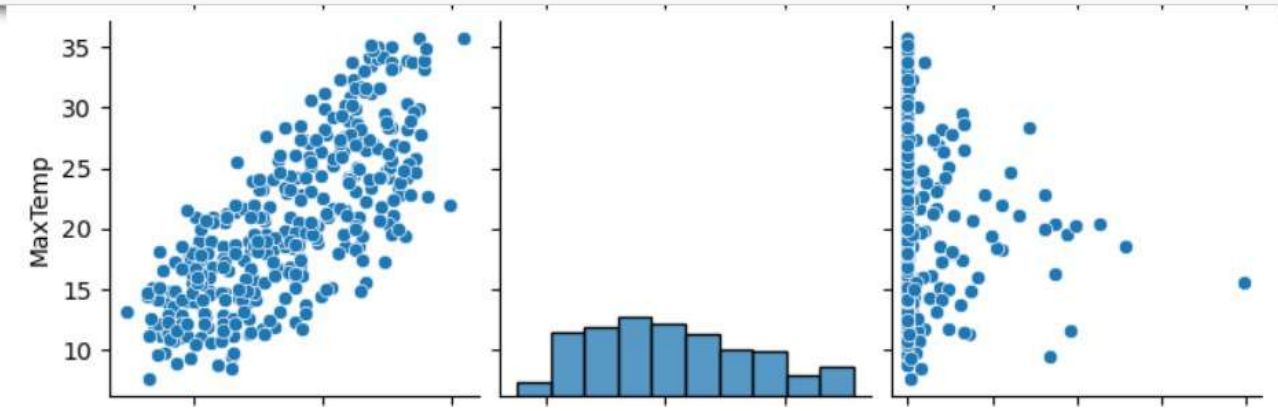
```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error

# Step 1: Load the Data
df = pd.read_csv('C:\\Users\\Teja\\Desktop\\work\\weather.csv')

# Step 2: Data Exploration
print(df.head())
print(df.info())
print(df.describe())

# Step 3: Data Visualization
sns.pairplot(df[['MinTemp', 'MaxTemp', 'Rainfall']])
plt.show()

# Step 4: Feature Engineering (if needed)
# No feature engineering needed for this example

# Step 5: Data Analysis (analyze each term)
# Example: Calculate average MaxTemp by month
# If you don't have a 'Date' column, you cannot create a 'Month' column.
```

```
# Instead, let's analyze MaxTemp directly or assume an artificial 'Month' column for this example.

# Assuming we add an artificial 'Month' column for demonstration
import numpy as np
df['Month'] = np.random.randint(1, 13, df.shape[0])
monthly_avg_max_temp = df.groupby('Month')['MaxTemp'].mean()

# Step 6: Data Visualization (Part 2)
plt.figure(figsize=(10, 5))
plt.plot(monthly_avg_max_temp.index, monthly_avg_max_temp.values, marker='o')
plt.xlabel('Month')
plt.ylabel('Average Max Temperature')
plt.title('Monthly Average Max Temperature')
plt.grid(True)
plt.show()

# Step 7: Advanced Analysis (e.g., predict Rainfall)
# Prepare the data for prediction
X = df[['MinTemp', 'MaxTemp']]
y = df['Rainfall']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Create and train a linear regression model
```

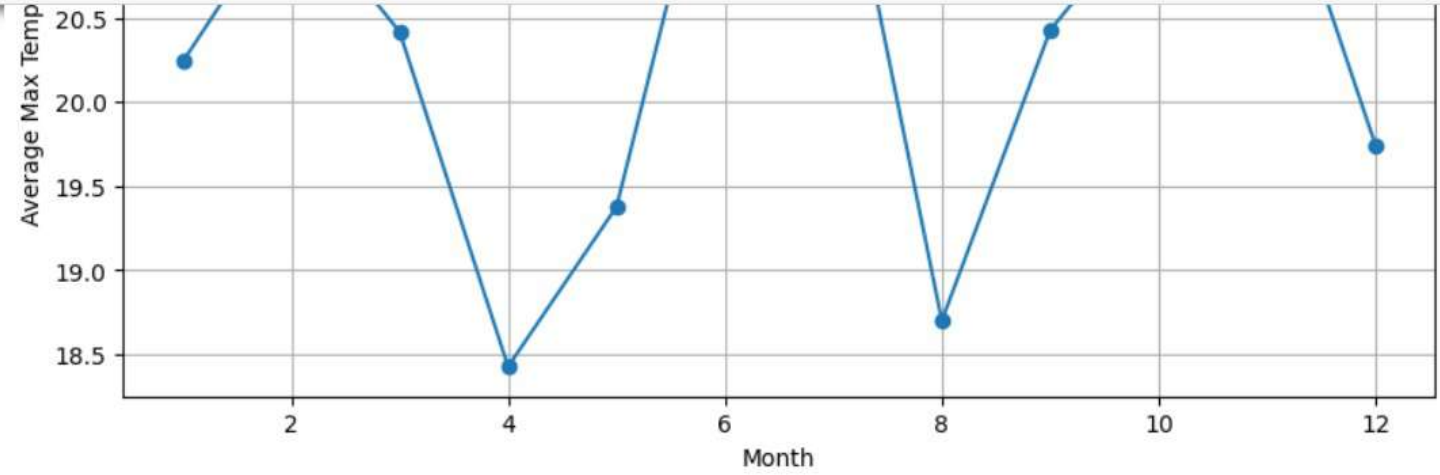
```
model = LinearRegression()
model.fit(X_train, y_train)

# Make predictions and calculate the Mean Squared Error
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print(f'Mean Squared Error for Rainfall Prediction: {mse}')

# Step 8: Conclusions and Insights (analyze each term)
# Example: Identify the highest and lowest rainfall months
monthly_avg_rainfall = df.groupby('Month')['Rainfall'].mean()
highest_rainfall_month = monthly_avg_rainfall.idxmax()
lowest_rainfall_month = monthly_avg_rainfall.idxmin()
print(f'Highest rainfall month: {highest_rainfall_month}, Lowest rainfall month:
{lowest_rainfall_month}')
```


MaxTemp
35
30
25
20
15
10

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	\
0	8.0	24.3	0.0	3.4	6.3	NW	
1	14.0	26.9	3.6	4.4	9.7	ENE	
2	13.7	23.4	3.6	5.8	3.3	NW	
3	13.3	15.5	39.8	7.2	9.1	NW	
4	7.6	16.1	2.8	5.6	10.6	SSE	

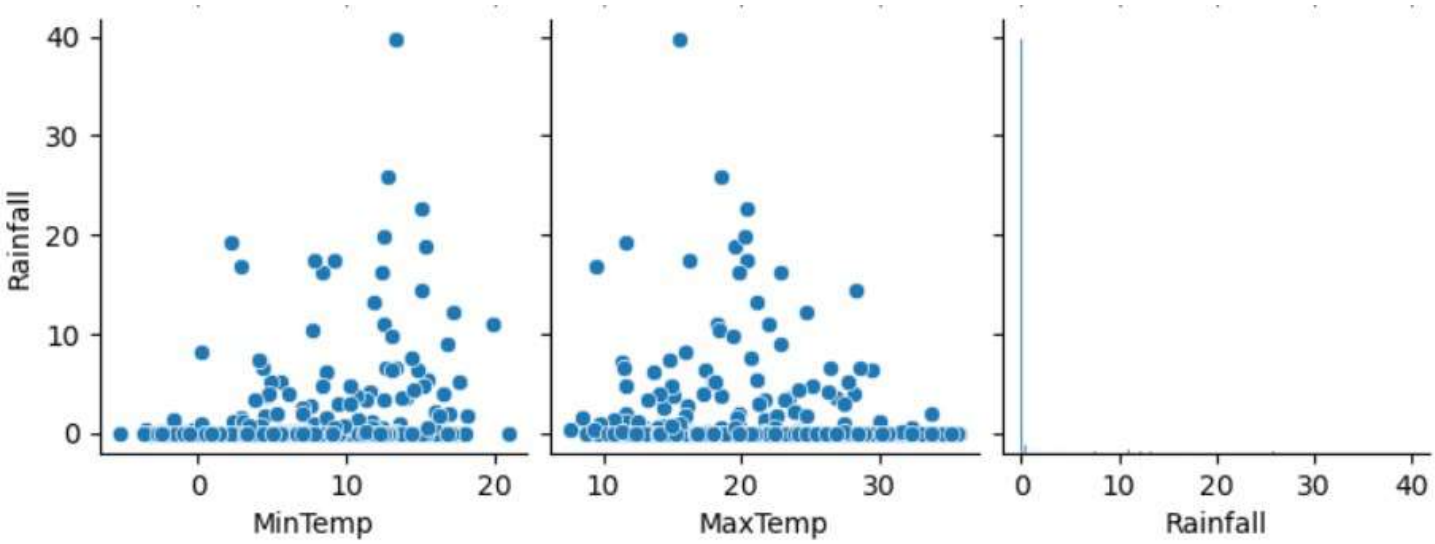
	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	...	Humidity3pm	\
0	30.0	SW	NW	6.0	...	29	
1	39.0	E	W	4.0	...	36	
2	85.0	N	NNE	6.0	...	69	
3	54.0	WNW	W	30.0	...	56	
4	50.0	SSE	ESE	20.0	...	49	

	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	\
0	1019.7	1015.0	7	7	14.4	23.6	No	
1	1012.4	1008.4	5	3	17.5	25.7	Yes	
2	1009.5	1007.2	8	7	15.4	20.2	Yes	
3	1005.5	1007.0	2	7	13.5	14.1	Yes	



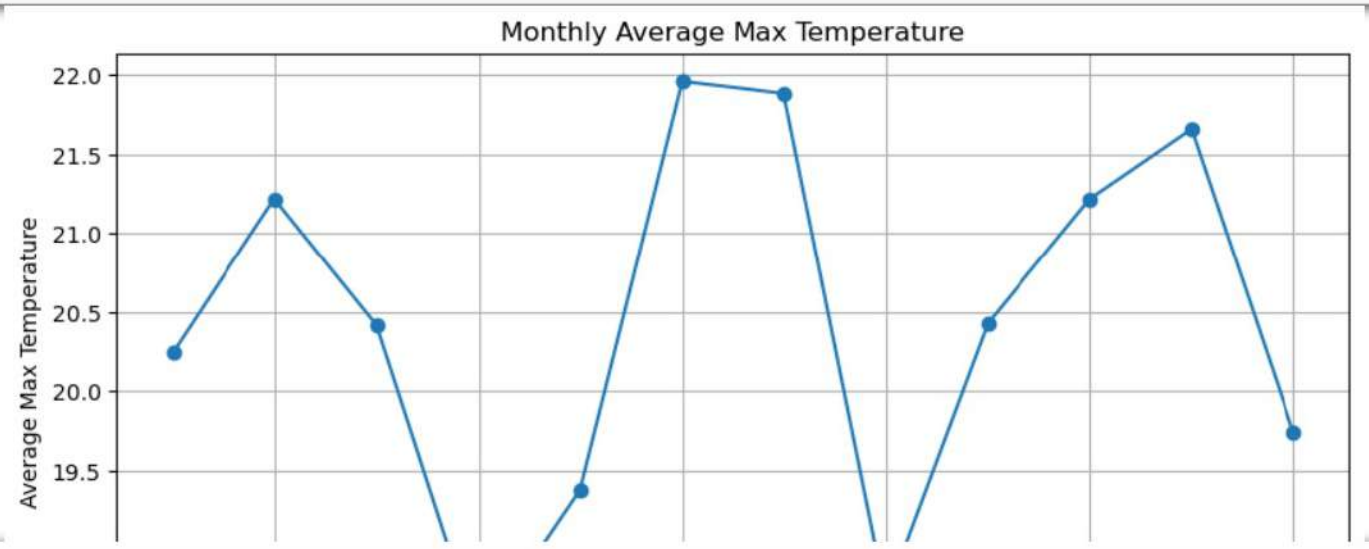

```
Mean Squared Error for Rainfall Prediction: 37.076845600582615
Highest rainfall month: 11, Lowest rainfall month: 7
```

0	MinTemp	366 non-null	float64
1	MaxTemp	366 non-null	float64
2	Rainfall	366 non-null	float64
3	Evaporation	366 non-null	float64
4	Sunshine	363 non-null	float64
5	WindGustDir	363 non-null	object
6	WindGustSpeed	364 non-null	float64
7	WindDir9am	335 non-null	object
8	WindDir3pm	365 non-null	object
9	WindSpeed9am	359 non-null	float64
10	WindSpeed3pm	366 non-null	int64
11	Humidity9am	366 non-null	int64
12	Humidity3pm	366 non-null	int64
13	Pressure9am	366 non-null	float64
14	Pressure3pm	366 non-null	float64
15	Cloud9am	366 non-null	int64
16	Cloud3pm	366 non-null	int64
17	Temp9am	366 non-null	float64



Monthly Average Max Temperature

```
 18  Temp3pm         366 non-null    float64
 19  RainToday       366 non-null    object
 20  RISK_MM         366 non-null    float64
 21  RainTomorrow    366 non-null    object
dtypes: float64(12), int64(5), object(5)
memory usage: 63.0+ KB
None
          MinTemp     MaxTemp    Rainfall  Evaporation    Sunshine  \
count  366.000000  366.000000  366.000000   366.000000  363.000000
mean     7.265574   20.550273    1.428415     4.521858    7.909366
std      6.025800    6.690516    4.225800     2.669383    3.481517
min     -5.300000    7.600000    0.000000     0.200000    0.000000
25%      2.300000   15.025000    0.000000     2.200000    5.950000
50%      7.450000   19.650000    0.000000     4.200000    8.600000
75%     12.500000   25.500000    0.200000     6.400000   10.500000
max     20.900000   35.800000   39.800000    13.800000   13.600000
```


Monthly Average Max Temperature
Average Max Temperature
22.0
21.5
21.0
20.5
20.0
19.5

	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	\
count	366.000000	366.000000	366.000000	366.000000	366.000000	
mean	1019.709016	1016.810383	3.890710	4.024590	12.358470	
std	6.686212	6.469422	2.956131	2.666268	5.630832	
min	996.500000	996.800000	0.000000	0.000000	0.100000	
25%	1015.350000	1012.800000	1.000000	1.000000	7.625000	
50%	1020.150000	1017.400000	3.500000	4.000000	12.550000	
75%	1024.475000	1021.475000	7.000000	7.000000	17.000000	
max	1035.700000	1033.200000	8.000000	8.000000	24.700000	

	Temp3pm	RISK_MM
count	366.000000	366.000000
mean	19.230874	1.428415
std	6.640346	4.225800
min	5.100000	0.000000
25%	14.150000	0.000000
50%	18.550000	0.000000
75%	24.000000	0.200000