WEATHER ANALYSIS

- Data
 Preparation
 with Python
- 2. Advanced
 Analysis with
 Power BI or
 Tableau
- 3. Correlation and Regression Analysis
- 4. Documentation

Weather Analysis

Project Requirements:

- 1. Data Preparation with Python:
- Use Python to clean and preprocess the weather dataset.
- Handle missing values, outliers, and any other inconsistencies in the data.
- 2. Advanced Analysis with Power BI or Tableau:
- Leverage Power BI or Tableau for advanced data analysis on the weather dataset.
- Create interactive dashboards and visualizations to highlight trends and patterns.
- 3. Correlation and Regression Analysis:
- Perform correlation analysis to identify relationships between different weather parameters.
- Implement regression analysis to predict one weather parameter based on others.
- 4. Documentation:
- Document your data preparation and analysis steps.
- Explain the insights derived from the advanced analysis.

WEATHER FORECASTING APPLICATION USING PYTHON

ABSTRACT

Weather forecasting is the application of science and technology to predict the state of the atmosphere for a given location. Ancient weather forecasting methods usually relied on observed patterns of events, also termed pattern recognition.

Here this system will predict weather based on parameters such as temperature, humidity and wind. Weather forecasting system takes parameters such as temperature, humidity, and wind and will forecast weather based on previous record therefore this prediction will prove reliable. This system can be used in Air Traffic, Marine, Agriculture, Forestry, Military, and Navy etc.

KEYWORDS

- ✓ Weather forecasting
- √ temperature
- ✓ short range
- ✓ challenging task
- ✓ forecasting

FEATURES WEATHER FORECAST

- ✓ Time to time update weather
- ✓ Temperature Update
- ✓ change weather in every hours as according to weather changes.
- ✓ provide accurate data information about weather.

PROCESSING AND CLEANING DATA FOR TABLEAU

In order to create effective visualizations in Tableau, it is important to ensure that the data is in a format that can be easily consumed by the software.

1. Remove irrelevant or duplicate data:

Processing the data, it is important to remove any data that is irrelevant or duplicated. This can help you reduce the size of your data set.

2. Convert data types:

Tableau works best with certain data types, such as date/time, string, and numeric data. If your data set includes other data types, such as Boolean or object data.

3. Split and merge columns:

Sometimes the data set may contain columns that need to be split into multiple columns or merged into one. Tableau allows you to do this using the "Split" and "Merge" features.

4.Aggregate data:

Tableau works best with aggregated data, which means that is need to summarize the data by grouping it into categories or calculating summary statistics.

5.Visualization Examples

There are many different types of visualizations that can be used to display weather data in Tableau. Here are some examples of different types of visualizations,

- ✓ Line chart
- ✓ Heat map✓ Scatter plot
- ✓ Bar chart

SOURCE CODE:

import pandas as pd import json df=pd.read_csv("weather.csv") print(df)

OUTPUT

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	١
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	
361	9.0	30.7	0.0	7.6	12.1	NNW	
362	7.1	28.4	0.0	11.6	12.7	N	
363	12.5	19.9	0.0	8.4	5.3	ESE	
364	12.5	26.9	0.0	5.0	7.1	NW	
365	12.3	30.2	0.0	6.0	12.6	NW	

	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	
361	76.0	SSE	NW	7.0	 15	
362	48.0	NNW	NNW	2.0	 22	
363	43.0	ENE	ENE	11.0	 47	
364	46.0	SSW	WNW	6.0	 39	
365	78.0	NW	WNW	31.0	 13	

	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	\
0	1019.7	1015.0	7	7	14.4	23.6	
1	1012.4	1008.4	5	3	17.5	25.7	
2	1009.5	1007.2	8	7	15.4	20.2	
3	1005.5	1007.0	2	7	13.5	14.1	
4	1018.3	1018.5	7	7	11.1	15.4	
361	1016.1	1010.8	1	3	20.4	30.0	
362	1020.0	1016.9	0	1	17.2	28.2	
363	1024.0	1022.8	3	2	14.5	18.3	
364	1021.0	1016.2	6	7	15.8	25.9	
365	1009.6	1009.2	1	1	23.8	28.6	

	RainToday	RISK_MM	${\tt RainTomorrow}$
0	No	3.6	Yes
1	Yes	3.6	Yes
2	Yes	39.8	Yes
3	Yes	2.8	Yes
4	Yes	0.0	No
361	No	0.0	No
362	No	0.0	No
363	No	0.0	No
364	No	0.0	No
365	No	0.0	No

[366 rows x 22 columns]

```
from IPython.core.interactiveshell import InteractiveShell
InteractiveShell.ast_node_interactivity = "all"
df.describe()
df[df['MaxTemp'] <= 15]
df.head()
df[df['Sunshine'] == 8.600000]
df.info()
df.dtypes
df.columns
df.describe().transpose()
df.isnull().sum()
df=df.drop_duplicates()
df.info()
cols=df.columns.tolist()
for i, col in enumerate(cols):
  if '(' in col:
    cols[i] = col + ")"
     df.cols= cols
  print(df.columns)
```

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustSpeed
count	366.000000	366.000000	366.000000	366.000000	363.000000	364.000000
mean	7.265574	20.550273	1.428415	4.521858	7.909366	39.840659
std	6.025800	6.690516	4.225800	2.669383	3.481517	13.059807
min	-5.300000	7.600000	0.000000	0.200000	0.000000	13.000000
25%	2.300000	15.025000	0.000000	2.200000	5.950000	31.000000
50%	7.450000	19.650000	0.000000	4.200000	8.600000	39.000000
75 %	12.500000	25.500000	0.200000	6.400000	10.500000	46.000000
max	20.900000	35.800000	39.800000	13.800000	13.600000	98.000000

WindSpeed9am	WindSpeed3pm	Humidity9am	Humidity3pm	Pressure9am	Pressure3pm
359.000000	366.000000	366.000000	366.000000	366.000000	366.000000
9.651811	17.986339	72.035519	44.519126	1019.709016	1016.810383
7.951929	8.856997	13.137058	16.850947	6.686212	6.469422
0.000000	0.000000	36.000000	13.000000	996.500000	996.800000
6.000000	11.000000	64.000000	32.250000	1015.350000	1012.800000
7.000000	17.000000	72.000000	43.000000	1020.150000	1017.400000
13.000000	24.000000	81.000000	55.000000	1024.475000	1021.475000
41.000000	52.000000	99.000000	96.000000	1035.700000	1033.200000

Cloud9am	Cloud3pm	Temp9am	Temp3pm	RISK_MM
366.000000	366.000000	366.000000	366.000000	366.000000
3.890710	4.024590	12.358470	19.230874	1.428415
2.956131	2.666268	5.630832	6.640346	4.225800
0.000000	0.000000	0.100000	5.100000	0.000000
1.000000	1.000000	7.625000	14.150000	0.000000
3.500000	4.000000	12.550000	18.550000	0.000000
7.000000	7.000000	17.000000	24.000000	0.200000
8.000000	8.000000	24.700000	34.500000	39.800000

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	
141	13.0	14.8	0.0	8.2	0.0	SE	30.0	ESE	ESE	9.0	
154	7.0	14.3	2.6	9.6	9.7	WNW	63.0	NW	W	26.0	
179	4.3	11.3	7.2	4.4	5.6	W	57.0	WNW	WNW	26.0	
180	-2.1	13.8	0.2	1.8	9.5	NNW	22.0	NaN	NNW	0.0	
181	-1.8	14.8	0.0	1.4	7.0	N	28.0	E	N	2.0	
309	5.8	12.4	0.0	1.6	0.0	SE	26.0	SSE	ESE	13.0	
313	-3.7	14.7	0.0	3.4	10.9	SSE	43.0	SE	N	9.0	
320	3.9	13.2	3.4	6.6	11.0	WNW	65.0	WNW	WNW	26.0	
321	0.7	14.1	0.0	5.6	9.0	ENE	20.0	SSW	NNW	6.0	
356	3.4	15.0	0.8	4.8	11.7	S	70.0	S	S	35.0	

92 rows × 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
62	1017.4	1019.3	7	8	13.3	13.9	No	0.0	No
40	1010.3	1012.7	1	3	9.1	13.7	Yes	0.0	No
60	1006.3	1008.1	3	6	7.1	8.8	Yes	0.2	No
40	1020.6	1019.6	0	1	6.3	13.2	No	0.0	No
40	1024.2	1020.5	1	7	5.3	13.9	No	0.0	No
73	1027.3	1023.5	8	8	8.7	10.8	No	0.0	No
25	1025.8	1020.9	0	0	5.5	13.4	No	0.0	No
33	1017.0	1017.6	3	1	7.1	12.2	Yes	0.0	No
43	1026.7	1022.1	7	1	7.4	13.7	No	0.0	No
24	1023.4	1023.1	1	5	8.3	14.3	No	0.0	No

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	
0	8.0	24.3	0.0	3.4	6.3	NW	30.0	SW	NW	6.0	
1	14.0	26.9	3.6	4.4	9.7	ENE	39.0	E	W	4.0	
2	13.7	23.4	3.6	5.8	3.3	NW	85.0	N	NNE	6.0	
3	13.3	15.5	39.8	7.2	9.1	NW	54.0	WNW	W	30.0	
4	7.6	16.1	2.8	5.6	10.6	SSE	50.0	SSE	ESE	20.0	

5 rows × 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
29	1019.7	1015.0	7	7	14.4	23.6	No	3.6	Yes
36	1012.4	1008.4	5	3	17.5	25.7	Yes	3.6	Yes
69	1009.5	1007.2	8	7	15.4	20.2	Yes	39.8	Yes
56	1005.5	1007.0	2	7	13.5	14.1	Yes	2.8	Yes
49	1018.3	1018.5	7	7	11.1	15.4	Yes	0.0	No

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm
113	12.7	28.6	6.6	3.2	8.6	W	50.0	NaN	W
211	-0.1	18.0	0.0	2.0	8.6	NaN	NaN	N	NNW
271	0.8	12.2	0.0	1.8	8.6	NaN	NaN	S	S
311	-0.9	16.7	0.0	2.8	8.6	NW	35.0	WNW	NW
322	1.1	18.0	0.0	1.6	8.6	NNW	39.0	NNW	NNW
352	6.7	24.7	0.0	5.4	8.6	NW	43.0	N	NW

6 rows × 22 columns

WindSpeed9am	 Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
NaN	 42	1008.3	1002.3	5	5	16.5	27.4	Yes	0.0	No
2.0	 46	1033.2	1028.7	1	1	5.6	17.4	No	0.0	No
19.0	 49	1020.9	1016.8	3	1	6.5	11.2	No	0.0	No
7.0	 33	1018.1	1013.9	7	7	6.4	15.5	No	1.0	No
2.0	 36	1021.0	1015.6	4	1	7.0	17.8	No	0.0	No
4.0	 31	1025.7	1022.2	1	7	12.7	23.7	No	0.0	No

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 366 entries, 0 to 365
Data columns (total 22 columns):

#	Column	Non-Null Count	Dtype
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
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

```
MinTemp
                 float64
MaxTemp
                 float64
Rainfall
                 float64
Evaporation
                 float64
Sunshine
                 float64
WindGustDir
                  object
                 float64
WindGustSpeed
WindDir9am
                  object
WindDir3pm
                  object
WindSpeed9am
                 float64
WindSpeed3pm
                   int64
Humidity9am
                   int64
Humidity3pm
                   int64
Pressure9am
                 float64
Pressure3pm
                 float64
Cloud9am
                   int64
Cloud3pm
                   int64
Temp9am
                 float64
Temp3pm
                 float64
RainToday
                  object
                 float64
RISK MM
RainTomorrow
                  object
dtype: object
```

	count	mean	std	min	25%	50%	75%	max
MinTemp	366.0	7.265574	6.025800	-5.3	2.300	7.45	12.500	20.9
MaxTemp	366.0	20.550273	6.690516	7.6	15.025	19.65	25.500	35.8
Rainfall	366.0	1.428415	4.225800	0.0	0.000	0.00	0.200	39.8
Evaporation	366.0	4.521858	2.669383	0.2	2.200	4.20	6.400	13.8
Sunshine	363.0	7.909366	3.481517	0.0	5.950	8.60	10.500	13.6
WindGustSpeed	364.0	39.840659	13.059807	13.0	31.000	39.00	46.000	98.0
WindSpeed9am	359.0	9.651811	7.951929	0.0	6.000	7.00	13.000	41.0
WindSpeed3pm	366.0	17.986339	8.856997	0.0	11.000	17.00	24.000	52.0
Humidity9am	366.0	72.035519	13.137058	36.0	64.000	72.00	81.000	99.0
Humidity3pm	366.0	44.519126	16.850947	13.0	32.250	43.00	55.000	96.0
Pressure9am	366.0	1019.709016	6.686212	996.5	1015.350	1020.15	1024.475	1035.7
Pressure3pm	366.0	1016.810383	6.469422	996.8	1012.800	1017.40	1021.475	1033.2
Cloud9am	366.0	3.890710	2.956131	0.0	1.000	3.50	7.000	8.0
Cloud3pm	366.0	4.024590	2.666268	0.0	1.000	4.00	7.000	8.0
Temp9am	366.0	12.358470	5.630832	0.1	7.625	12.55	17.000	24.7
Temp3pm	366.0	19.230874	6.640346	5.1	14.150	18.55	24.000	34.5
RISK_MM	366.0	1.428415	4.225800	0.0	0.000	0.00	0.200	39.8

MinTemp	0
MaxTemp	0
Rainfall	0
Evaporation	0
Sunshine	3
WindGustDir	3
WindGustSpeed	2
WindDir9am	31
WindDir3pm	1
WindSpeed9am	7
WindSpeed3pm	0
Humidity9am	0
Humidity3pm	0
Pressure9am	0
Pressure3pm	0
Cloud9am	0
Cloud3pm	0
Temp9am	0
Temp3pm	0
RainToday	0
RISK_MM	0
RainTomorrow	0
dtype: int64	

<class 'pandas.core.frame.DataFrame'> RangeIndex: 366 entries, 0 to 365 Data columns (total 22 columns):

#	Column	Non-Null Count	Dtype
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
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
dtvp	es: float64(12)	, int64(5), obje	ect(5)

dtypes: float64(12), int64(5), object(5)

memory usage: 63.0+ KB

```
['MinTemp',
 'MaxTemp',
 'Rainfall',
 'Evaporation',
 'Sunshine',
 'WindGustDir',
 'WindGustSpeed',
 'WindDir9am',
 'WindDir3pm',
 'WindSpeed9am',
 'WindSpeed3pm',
 'Humidity9am',
 'Humidity3pm',
 'Pressure9am',
 'Pressure3pm',
 'Cloud9am',
 'Cloud3pm',
 'Temp9am',
 'Temp3pm',
 'RainToday',
 'RISK_MM',
 'RainTomorrow']
```

```
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
 Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
 Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
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        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
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        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
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Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
```

```
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
         'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
         'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm'
         'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
         'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
        dtype='object')
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         'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
         'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
         'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
         'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
        dtype='object')
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         'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
         'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
         'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
         'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
        dtype='object')
 Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
         'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
         'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
         'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
        dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm', 'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
       dtype='object')
```

```
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtvpe='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
       'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
       'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
       'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
       'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
df.columns
df.to_csv('cln_weather_data.csv', encoding='utf-8', index=False)
df['MaxTemp'].value_counts()
Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
        'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm',
        'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
        'Pressure9am', 'Pressure3pm', 'Cloud9am', 'Cloud3pm', 'Temp9am',
        'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
      dtype='object')
                MaxTemp
                14.8
                20.9
                       5
                12.2
                15.5
                        5
                11.6
                       5
                      1
                17.1
                30.3
                        1
                        1
                35.8
                21.2
                30.7
                Name: count, Length: 187, dtype: int64
```

$df[df['Rainfall'] \le 3.6]$

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	•••
0	8.0	24.3	0.0	3.4	6.3	NW	30.0	SW	NW	6.0	
1	14.0	26.9	3.6	4.4	9.7	ENE	39.0	Е	W	4.0	
2	13.7	23.4	3.6	5.8	3.3	NW	85.0	N	NNE	6.0	
4	7.6	16.1	2.8	5.6	10.6	SSE	50.0	SSE	ESE	20.0	
5	6.2	16.9	0.0	5.8	8.2	SE	44.0	SE	E	20.0	
361	9.0	30.7	0.0	7.6	12.1	NNW	76.0	SSE	NW	7.0	
362	7.1	28.4	0.0	11.6	12.7	N	48.0	NNW	NNW	2.0	
363	12.5	19.9	0.0	8.4	5.3	ESE	43.0	ENE	ENE	11.0	
364	12.5	26.9	0.0	5.0	7.1	NW	46.0	SSW	WNW	6.0	
365	12.3	30.2	0.0	6.0	12.6	NW	78.0	NW	WNW	31.0	

323 rows × 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
29	1019.7	1015.0	7	7	14.4	23.6	No	3.6	Yes
36	1012.4	1008.4	5	3	17.5	25.7	Yes	3.6	Yes
69	1009.5	1007.2	8	7	15.4	20.2	Yes	39.8	Yes
49	1018.3	1018.5	7	7	11.1	15.4	Yes	0.0	No
57	1023.8	1021.7	7	5	10.9	14.8	No	0.2	No
15	1016.1	1010.8	1	3	20.4	30.0	No	0.0	No
22	1020.0	1016.9	0	1	17.2	28.2	No	0.0	No
47	1024.0	1022.8	3	2	14.5	18.3	No	0.0	No
39	1021.0	1016.2	6	7	15.8	25.9	No	0.0	No
13	1009.6	1009.2	1	1	23.8	28.6	No	0.0	No

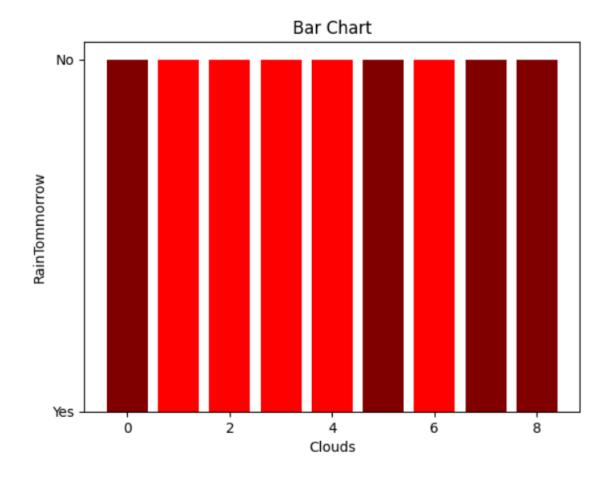
$df.group by ('WindGustDir').get_group ('NW')$

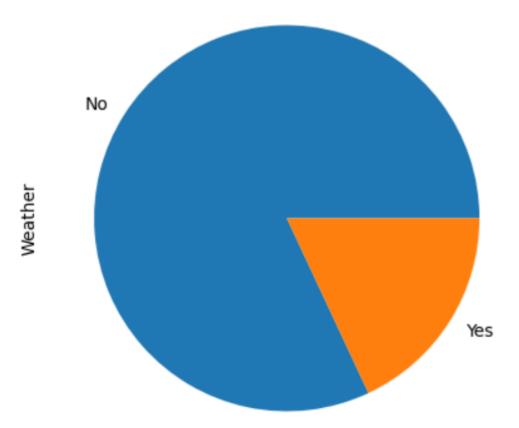
	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	
0	8.0	24.3	0.0	3.4	6.3	NW	30.0	SW	NW	6.0	
2	13.7	23.4	3.6	5.8	3.3	NW	85.0	N	NNE	6.0	
3	13.3	15.5	39.8	7.2	9.1	NW	54.0	WNW	W	30.0	
13	12.1	30.9	0.0	6.2	12.4	NW	44.0	WNW	W	7.0	
14	10.1	31.2	0.0	8.8	13.1	NW	41.0	S	W	6.0	
352	6.7	24.7	0.0	5.4	8.6	NW	43.0	N	NW	4.0	
353	8.3	28.5	0.0	5.8	9.8	NW	46.0	W	NW	2.0	
354	11.3	27.4	0.2	7.6	12.1	NW	52.0	SE	NW	6.0	
364	12.5	26.9	0.0	5.0	7.1	NW	46.0	SSW	WNW	6.0	
365	12.3	30.2	0.0	6.0	12.6	NW	78.0	NW	WNW	31.0	

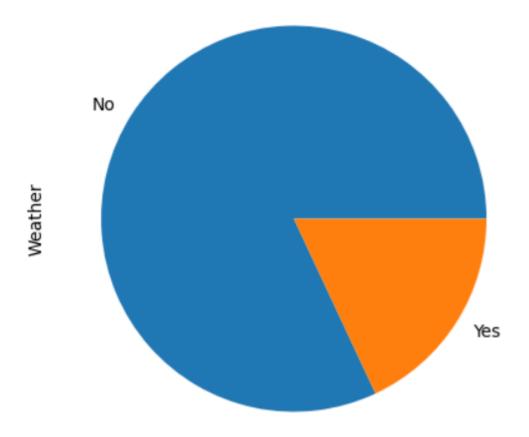
73 rows \times 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
29	1019.7	1015.0	7	7	14.4	23.6	No	3.6	Yes
69	1009.5	1007.2	8	7	15.4	20.2	Yes	39.8	Yes
56	1005.5	1007.0	2	7	13.5	14.1	Yes	2.8	Yes
20	1017.3	1013.1	1	4	19.7	30.7	No	0.0	No
16	1018.2	1013.7	0	1	18.7	30.4	No	0.0	No
31	1025.7	1022.2	1	7	12.7	23.7	No	0.0	No
30	1024.1	1019.8	1	6	16.8	27.4	No	0.2	No
20	1021.4	1017.5	1	1	16.4	26.3	No	0.0	No
39	1021.0	1016.2	6	7	15.8	25.9	No	0.0	No
13	1009.6	1009.2	1	1	23.8	28.6	No	0.0	No

```
import matplotlib
import matplotlib.pyplot as plt
fig = plt.figure()
\#ax = fig.add\_axes([0,1,1,1])
plt.bar(df['Cloud3pm'],df['RainTomorrow'],color = ['maroon','red'])
plt.title('Bar Chart')
plt.xlabel("Clouds")
plt.ylabel("RainTommorrow")
plt.show()
<BarContainer object of 366 artists>
Text(0.5, 1.0, 'Bar Chart')
Text(0.5, 0, 'Clouds')
Text(0, 0.5, 'RainTommorrow')
weather = dict(df['RainTomorrow'].value_counts())
series = pd.Series(weather.values(), index=weather.keys(), name="Weather")
series.plot.pie(figsize=(5, 5))
<Axes: ylabel='Weather'>
weather = dict(df['RainToday'].value_counts())
series = pd.Series(weather.values(), index=weather.keys(), name="Weather")
series.plot.pie(figsize=(5, 5))
<Axes: ylabel='Weather'>
```







 $df[(df['WindDir9am'] == 'E')\& \ (df['WindDir3pm'] == 'E')]$

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	
25	15.6	26.9	0.0	6.8	8.9	Е	41.0	Е	Е	6.0	
26	13.3	22.2	0.2	6.6	2.3	ENE	39.0	Е	Е	20.0	

2 rows × 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
48	1019.7	1016.5	2	4	19.8	25.1	No	0.2	No
55	1021.0	1018.6	7	7	16.5	21.2	No	0.0	No

df[(df['RainToday'] == 'No')& (df['WindDir3pm'] == 'E')]

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm	WindSpeed9am	1
5	6.2	16.9	0.0	5.8	8.2	SE	44.0	SE	Е	20.0	
7	8.3	17.0	0.0	5.6	4.6	Е	41.0	SE	Е	11.0	
21	16.4	19.4	0.4	9.2	0.0	Е	26.0	ENE	Е	6.0	
25	15.6	26.9	0.0	6.8	8.9	Е	41.0	Е	Е	6.0	
26	13.3	22.2	0.2	6.6	2.3	ENE	39.0	Е	Е	20.0	
34	14.5	21.8	0.0	8.4	9.8	ENE	43.0	ESE	Е	11.0	
40	11.8	18.5	0.6	4.8	2.3	ENE	35.0	ESE	Е	9.0	
41	11.7	21.5	0.0	4.2	7.3	ENE	41.0	ESE	Е	15.0	
42	9.6	20.3	0.0	5.0	3.6	SE	39.0	ESE	Е	22.0	
47	7.5	20.9	0.0	6.6	8.7	ENE	39.0	SE	Е	13.0	
65	12.1	27.5	0.0	7.4	11.7	NE	35.0	SSE	Е	7.0	
83	9.9	24.4	0.0	5.8	10.8	NE	28.0	SE	Е	7.0	
106	10.8	25.2	0.0	5.6	12.6	ENE	35.0	SE	Е	7.0	
107	11.2	26.1	0.0	7.2	12.6	ENE	39.0	SE	Е	7.0	
173	5.6	18.9	0.0	3.8	8.0	SE	33.0	SSE	Е	13.0	
203	2.4	14.7	0.0	2.6	9.2	ESE	20.0	SSE	Е	7.0	
266	-1.0	12.2	0.0	1.6	8.4	ESE	30.0	SE	Е	7.0	

17 rows × 22 columns

Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am	Temp3pm	RainToday	RISK_MM	RainTomorrow
57	1023.8	1021.7	7	5	10.9	14.8	No	0.2	No
57	1026.2	1024.2	6	7	12.1	15.5	No	0.0	No
72	1010.7	1008.9	8	8	16.5	18.3	No	25.8	Yes
48	1019.7	1016.5	2	4	19.8	25.1	No	0.2	No
55	1021.0	1018.6	7	7	16.5	21.2	No	0.0	No
63	1015.0	1015.3	7	1	18.6	20.5	No	0.0	No
63	1018.0	1017.4	8	6	13.8	16.4	No	0.0	No
51	1021.1	1019.6	7	5	15.7	19.4	No	0.0	No
49	1021.9	1019.3	7	7	15.8	19.2	No	0.0	No
43	1020.6	1018.3	6	7	15.3	19.8	No	0.0	No
40	1011.8	1007.4	7	2	18.6	26.8	No	0.0	No
38	1021.6	1019.2	1	4	16.2	22.7	No	0.0	No
32	1020.2	1016.7	1	1	15.0	24.3	No	0.0	No
40	1022.9	1020.0	1	1	15.3	24.7	No	0.0	No
49	1025.8	1023.5	7	6	12.5	17.7	No	0.0	No
43	1030.4	1026.0	1	1	5.9	13.8	No	0.0	No
51	1032.2	1027.4	3	5	3.7	10.8	No	0.0	No

CONCLUSION

We successfully predicted the rainfall using the linear regression but here this is not very accurate only sometimes any way it depends upon the climate changes to season to season.

Here we are taking only summer season weather data set it only useful to predict rainfall in summer season. Finally, it is agreed that we made an attempt on the following points:

- ✓ The description of the purpose the scope and applicability of this project.
- ✓ We specify the system's necessary specs as well as the actions that can be performed on these objects.
- ✓ We define the system's required specifications and the actions that can be taken on these objects.
- ✓ We comprehend the problem domain and create a system model that represents the operations that can be performed on the system.
- ✓ We went into great lengths about the features and processes, providing a lot of important information.
- ✓ We created the user interface as well as system security issues.
- ✓ Finally, the system is built and tested in accordance with the test cases.