Data Collection and Preprocessing Phase

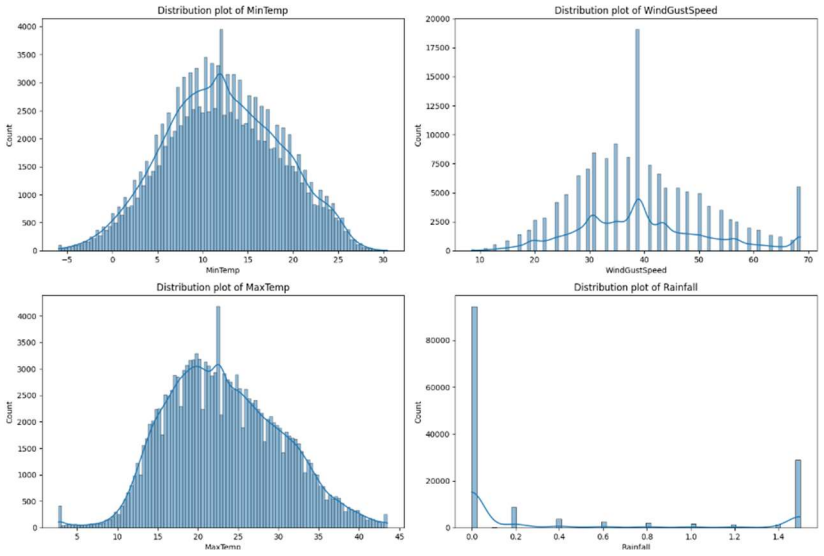
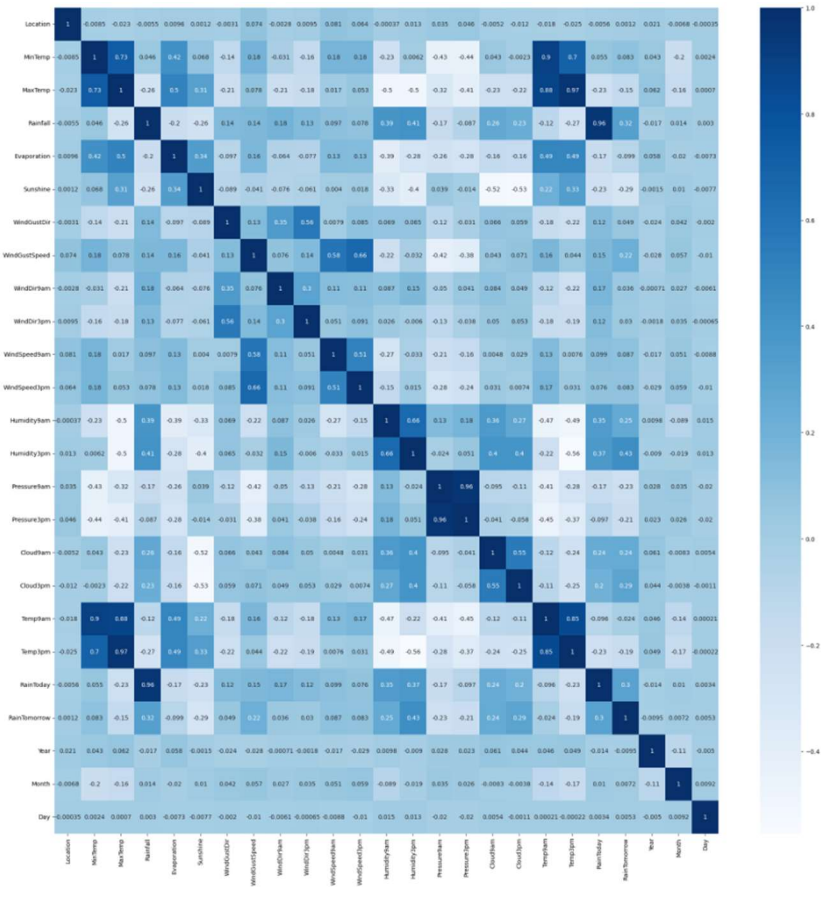
Date	23 September 2024
Team ID	LTVIP2024TMID24992
Project Title	Rainfall Prediction Using Machine Learning
Maximum Marks	6 Marks

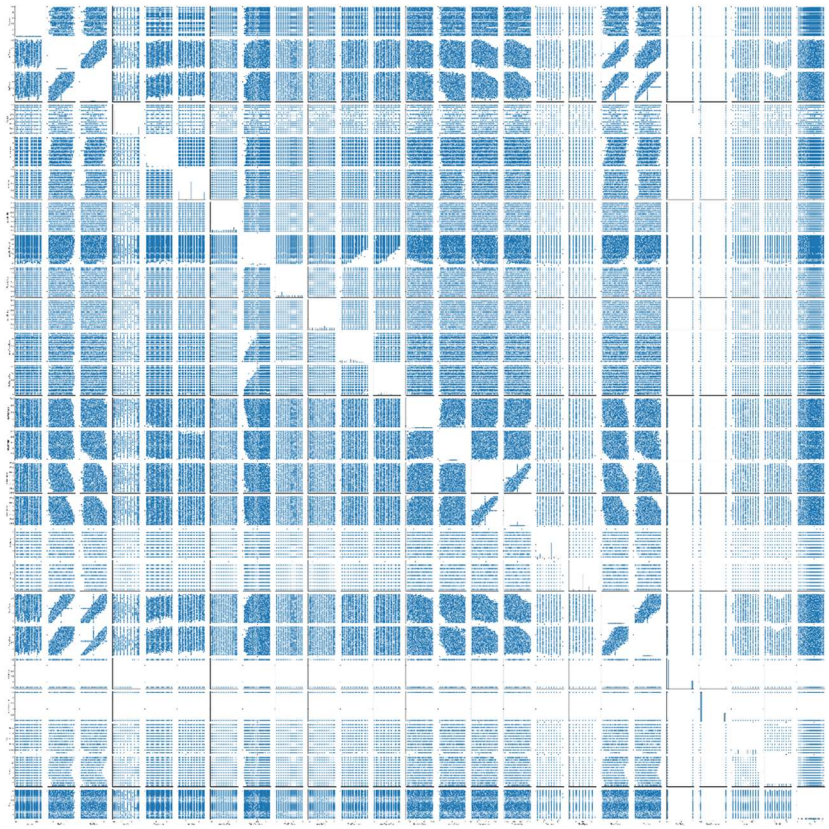
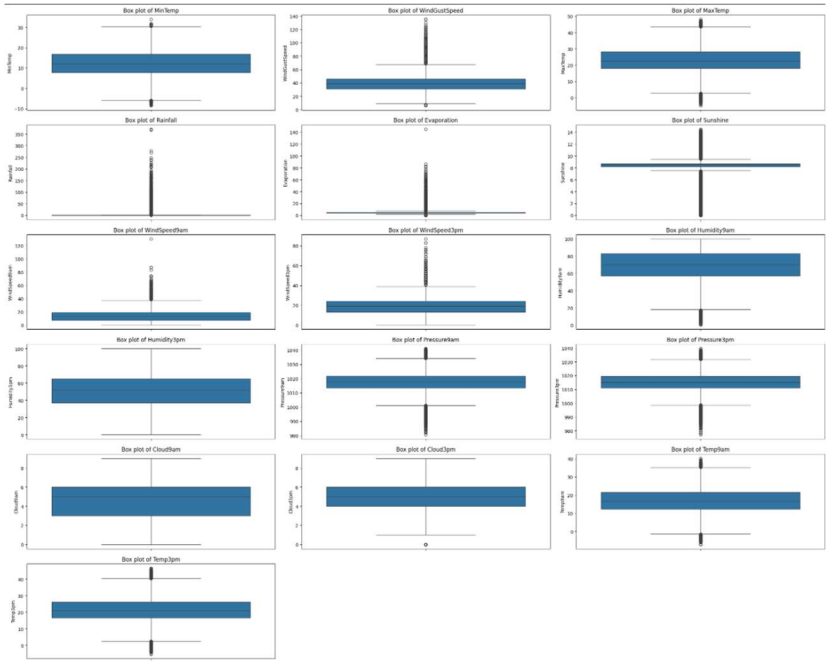
Data Exploration and Preprocessing Template

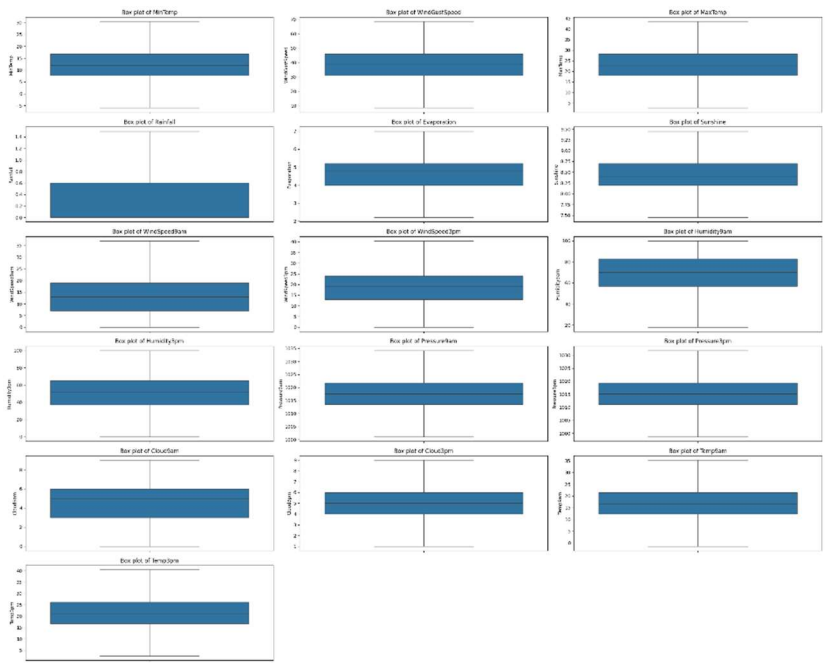
Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

Section	Description
Data Overview	Dimension: 145460 rows × 23 columns

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustSpeed	WindSpeed9am
count	143975.000000	144199.000000	142199.000000	82670.000000	75625.000000	135197.000000	143693.000000
mean	12.194034	23.221348	2.360918	5.468232	7.611178	40.035230	14.043426
std	6.398495	7.119049	8.478060	4.193704	3.785483	13.607062	8.915375
min	-8.500000	-4.800000	0.000000	0.000000	0.000000	6.000000	0.000000
25%	7.600000	17.900000	0.000000	2.600000	4.800000	31.000000	7.000000
50%	12.000000	22.600000	0.000000	4.800000	8.400000	39.000000	13.000000
75%	16.900000	28.200000	0.800000	7.400000	10.600000	48.000000	19.000000
max	33.900000	48.100000	371.000000	145.000000	14.500000	135.000000	130.000000

Univariate Analysis	
Bivariate Analysis	

Multivariate Analysis	
Outliers and Anomalies	Identification of outliers  Treatment of outliers.

<table>
<tr><td></td><td>  </td></tr>
<tr><td colspan="2">Data Preprocessing Code Screenshots</td></tr>
<tr><td>Loading Data</td><td>

```
[ ] df=pd.read_csv('/content/Weather.csv - Dataset.csv')
    df
```

	Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am
0	2008-12-01	Delhi	13.4	22.9	0.6	NaN	NaN	W	44.0	W
1	2008-12-02	Delhi	7.4	25.1	0.0	NaN	NaN	WNW	44.0	NNW
2	2008-12-03	Delhi	12.9	25.7	0.0	NaN	NaN	WSW	46.0	W
3	2008-12-04	Delhi	9.2	28.0	0.0	NaN	NaN	NE	24.0	SE
4	2008-12-05	Delhi	17.5	32.3	1.0	NaN	NaN	W	41.0	ENE
...	...	...	...	...	...	...	...	...	...	...
145455	2017-06-21	Uluru	2.8	23.4	0.0	NaN	NaN	E	31.0	SE
145456	2017-06-22	Uluru	3.6	25.3	0.0	NaN	NaN	NNW	22.0	SE
145457	2017-06-23	Uluru	5.4	26.9	0.0	NaN	NaN	N	37.0	SE
145458	2017-06-24	Uluru	7.8	27.0	0.0	NaN	NaN	SE	28.0	SSE
145459	2017-06-25	Uluru	14.9	NaN	0.0	NaN	NaN	NaN	NaN	ESE

145460 rows × 23 columns

</td></tr>
<tr><td>Handling Missing Data</td><td>Identifying missing values</td></tr>
</table>

```
df.isnull().sum()
```

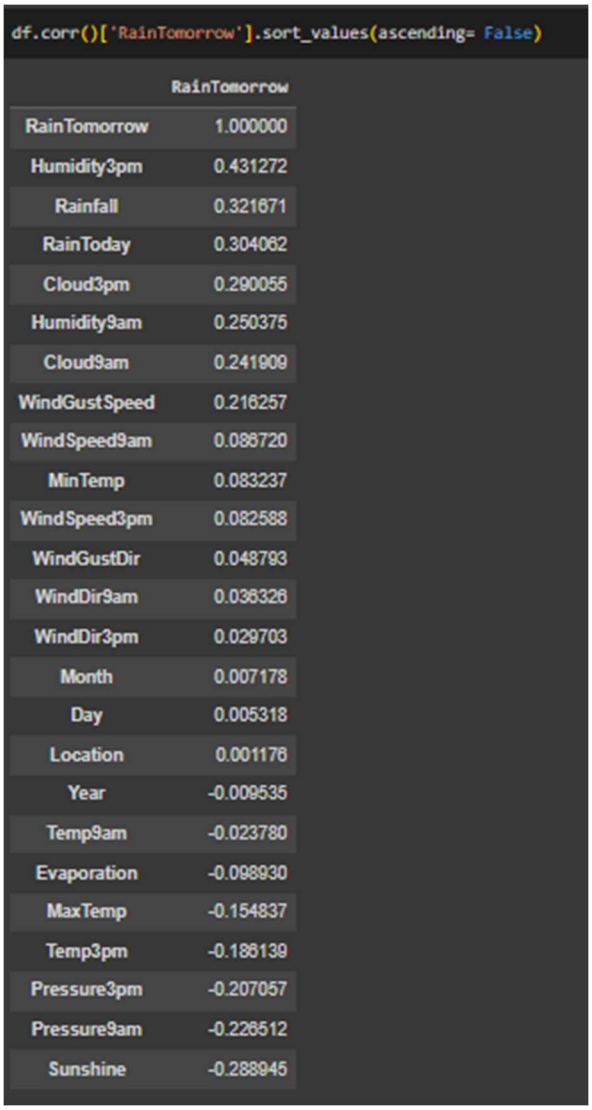
	0
Date	0
Location	0
MinTemp	1485
MaxTemp	1261
Rainfall	3261
Evaporation	62790
Sunshine	69835
WindGustDir	10326
WindGustSpeed	10263
WindDir9am	10566
WindDir3pm	4228
WindSpeed9am	1767
WindSpeed3pm	3062
Humidity9am	2654
Humidity3pm	4507
Pressure9am	15065
Pressure3pm	15028
Cloud9am	55888
Cloud3pm	59358
Temp9am	1767
Temp3pm	3609
RainToday	3261
RainTomorrow	3253

Handling missing values

```
df['MinTemp'].fillna(df['MinTemp'].median(), inplace = True)
df['MaxTemp'].fillna(df['MaxTemp'].median(), inplace = True)
df['Rainfall'].fillna(df['Rainfall'].median(), inplace = True)
df['Evaporation'].fillna(df['Evaporation'].median(), inplace = True)
df['Sunshine'].fillna(df['Sunshine'].median(), inplace = True)
df['WindGustSpeed'].fillna(df['WindGustSpeed'].median(), inplace = True)
df['WindSpeed9am'].fillna(df['WindSpeed9am'].median(), inplace = True)
df['WindSpeed3pm'].fillna(df['WindSpeed3pm'].median(), inplace = True)
df['Humidity9am'].fillna(df['Humidity9am'].median(), inplace = True)
df['Humidity3pm'].fillna(df['Humidity3pm'].median(), inplace = True)
df['Pressure9am'].fillna(df['Pressure9am'].median(), inplace = True)
df['Pressure3pm'].fillna(df['Pressure3pm'].median(), inplace = True)
df['Cloud9am'].fillna(df['Cloud9am'].median(), inplace = True)
df['Cloud3pm'].fillna(df['Cloud3pm'].median(), inplace = True)
df['Temp9am'].fillna(df['Temp9am'].median(), inplace = True)
df['Temp3pm'].fillna(df['Temp3pm'].median(), inplace = True)
```

	`df['WindDir9am'].fillna(df['WindDir9am'].mode()[0], inplace=True)`
	`df['WindDir3pm'].fillna(df['WindDir3pm'].mode()[0], inplace=True)`
	`df['RainToday'].fillna(df['RainToday'].mode()[0], inplace=True)`
	`df['RainTomorrow'].fillna(df['RainTomorrow'].mode()[0], inplace=True)`
	`df['WindGustDir'].fillna(df['WindGustDir'].mode()[0], inplace=True)`

Data Transformation	Encoding
	`le = LabelEncoder()`
	`df['WindDir9am'] = le.fit_transform(df['WindDir9am'])`
	`df['WindDir3pm'] = le.fit_transform(df['WindDir3pm'])`
	`df['RainToday'] = le.fit_transform(df['RainToday'])`
	`df['RainTomorrow'] = le.fit_transform(df['RainTomorrow'])`
	`df['Location'] = le.fit_transform(df['Location'])`
	`df['WindGustDir'] = le.fit_transform(df['WindGustDir'])`
	Scaling
	`x = df.loc[:, ['Humidity3pm', 'Rainfall', 'Cloud3pm', 'Humidity9am', 'Cloud9am', 'WindGustSpeed', 'WindSpeed9am', 'MinTemp', 'WindSpeed3pm','WindGustDir']]`
	`y = df['RainTomorrow']`
	Feature Scaling
	`#scaling the data`
	`sc = StandardScaler()`
	`x_sc = sc.fit_transform(x)`
	`x = pd.DataFrame(x_sc, columns=x.columns)`

Feature Engineering	`df.corr()['RainTomorrow'].sort_values(ascending= False)`  Selecting which are highly corelated to the target column and relatable.
Save Processed Data	-

	RainTomorrow
RainTomorrow	1.000000
Humidity3pm	0.431272
Rainfall	0.321671
RainToday	0.304062
Cloud3pm	0.290055
Humidity9am	0.250375
Cloud9am	0.241909
WindGustSpeed	0.216257
WindSpeed9am	0.088720
MinTemp	0.083237
WindSpeed3pm	0.082588
WindGustDir	0.048793
WindDir9am	0.036326
WindDir3pm	0.029703
Month	0.007178
Day	0.005318
Location	0.001176
Year	-0.009535
Temp9am	-0.023780
Evaporation	-0.098930
MaxTemp	-0.154837
Temp3pm	-0.188139
Pressure3pm	-0.207057
Pressure9am	-0.226512
Sunshine	-0.288945