



# WEATHERAUS

```
In [ ]: import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns
```

```
In [ ]: df5=pd.read_csv('weatherAUS.csv')
```

```
In [ ]: df5
```

```
Out[ ]:
```

	Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine
0	2008-12-01	Albury	13.4	22.9	0.6	NaN	NaN
1	2008-12-02	Albury	7.4	25.1	0.0	NaN	NaN
2	2008-12-03	Albury	12.9	25.7	0.0	NaN	NaN
3	2008-12-04	Albury	9.2	28.0	0.0	NaN	NaN
4	2008-12-05	Albury	17.5	32.3	1.0	NaN	NaN
...	...	...	...	...	...	...	...
94768	2011-01-26	Adelaide	16.7	29.6	0.4	NaN	12
94769	2011-01-27	Adelaide	19.5	32.0	0.0	12.4	9
94770	2011-01-28	Adelaide	15.8	31.8	0.0	8.0	12
94771	2011-01-29	Adelaide	18.5	37.7	0.0	NaN	12
94772	2011-01-30	Adelaide	25.8	42.5	0.0	NaN	12

94773 rows × 24 columns

```
In [ ]: df5.shape
```

```
Out[ ]: (94773, 24)
```

```
In [ ]: df5.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 94773 entries, 0 to 94772
Data columns (total 24 columns):
 #   Column            Non-Null Count  Dtype  
--- 
 0   Date              94773 non-null   object 
 1   Location          94773 non-null   object 
 2   MinTemp           94313 non-null   float64
 3   MaxTemp           94554 non-null   float64
 4   Rainfall          93718 non-null   float64
 5   Evaporation       52579 non-null   float64
 6   Sunshine          45783 non-null   float64
 7   WindGustDir       89081 non-null   object 
 8   WindGustSpeed     89089 non-null   float64
 9   WindDir9am        87119 non-null   object 
 10  WindDir3pm        92116 non-null   object 
 11  WindSpeed9am      93615 non-null   float64
 12  WindSpeed3pm      93014 non-null   float64
 13  Humidity9am       93482 non-null   float64
 14  Humidity3pm       93067 non-null   float64
 15  Pressure9am       85072 non-null   float64
 16  Pressure3pm       85119 non-null   float64
 17  Cloud9am          58619 non-null   float64
 18  Cloud3pm          57476 non-null   float64
 19  Temp9am           93982 non-null   float64
 20  Temp3pm           93514 non-null   float64
 21  RainToday          93717 non-null   object 
 22  RISK_MM            94772 non-null   float64
 23  RainTomorrow       94772 non-null   object 

dtypes: float64(17), object(7)
memory usage: 17.4+ MB
```

```
In [ ]: df5.isnull().sum()
```

Out[ ]:	0
<b>Date</b>	0
<b>Location</b>	0
<b>MinTemp</b>	460
<b>MaxTemp</b>	219
<b>Rainfall</b>	1055
<b>Evaporation</b>	42194
<b>Sunshine</b>	48990
<b>WindGustDir</b>	5692
<b>WindGustSpeed</b>	5684
<b>WindDir9am</b>	7654
<b>WindDir3pm</b>	2657
<b>WindSpeed9am</b>	1158
<b>WindSpeed3pm</b>	1759
<b>Humidity9am</b>	1291
<b>Humidity3pm</b>	1706
<b>Pressure9am</b>	9701
<b>Pressure3pm</b>	9654
<b>Cloud9am</b>	36154
<b>Cloud3pm</b>	37297
<b>Temp9am</b>	791
<b>Temp3pm</b>	1259
<b>RainToday</b>	1056
<b>RISK_MM</b>	1
<b>RainTomorrow</b>	1

**dtype:** int64

```
In [ ]: df5['MinTemp']=df5['MinTemp'].fillna(df5['MinTemp'].mean())
```

```
In [ ]: df5['MaxTemp']=df5['MaxTemp'].fillna(df5['MaxTemp'].mean())
```

```
In [ ]: df5['MinTemp'].isnull().sum()
```

```
Out[ ]: np.int64(0)
```

```
In [ ]: df5['MaxTemp'].isnull().sum()

Out[ ]: np.int64(0)

In [ ]: df5['Rainfall']=df5['Rainfall'].fillna(df5['Rainfall'].mean())

In [ ]: df5['Evaporation']=df5['Evaporation'].fillna(df5['Evaporation'].mean())

In [ ]: df5['Sunshine']=df5['Sunshine'].fillna(df5['Sunshine'].mean())

In [ ]: df5['WindGustDir'].unique()

Out[ ]: array(['W', 'WNW', 'WSW', 'NE', 'NNW', 'N', 'NNE', 'SW', 'ENE', 'SSE',
       'S', 'NW', 'SE', 'ESE', nan, 'E', 'SSW'], dtype=object)

In [ ]: df5['WindGustDir']=df5['WindGustDir'].fillna(np.random.choice(['W', 'WNW', 'WS',
       'S', 'NW', 'SE', 'ESE', 'E', 'SSW']))

In [ ]: df5['WindGustSpeed']=df5['WindGustSpeed'].fillna(df5['WindGustSpeed'].mean())

In [ ]: df5['WindDir9am'].unique()

Out[ ]: array(['W', 'NNW', 'SE', 'ENE', 'SW', 'SSE', 'S', 'NE', nan, 'SSW', 'N',
       'WSW', 'ESE', 'E', 'NW', 'WNW', 'NNE'], dtype=object)

In [ ]: df5['WindDir9am']=df5['WindDir9am'].fillna(np.random.choice(['W', 'NNW', 'SE',
       'WSW', 'ESE', 'E', 'NW', 'WNW', 'NNE']))

In [ ]: df5['WindDir3pm'].unique()

Out[ ]: array(['WNW', 'WSW', 'E', 'NW', 'W', 'SSE', 'ESE', 'ENE', 'NNW', 'SSW',
       'SW', 'SE', 'N', 'S', 'NNE', nan, 'NE'], dtype=object)

In [ ]: df5['WindDir3pm']=df5['WindDir3pm'].fillna(np.random.choice(['W', 'NNW', 'SE',
       'WSW', 'ESE', 'E', 'NW', 'WNW', 'NNE']))

In [ ]: df5['WindDir3pm'].isnull().sum()

Out[ ]: np.int64(0)

In [ ]: df5['WindSpeed9am']=df5['WindSpeed9am'].fillna(df5['WindSpeed9am'].mean())

In [ ]: df5['WindSpeed3pm']=df5['WindSpeed3pm'].fillna(df5['WindSpeed3pm'].mean())

In [ ]: df5['Humidity9am']=df5['Humidity9am'].fillna(df5['Humidity9am'].mean())

In [ ]: df5['Humidity3pm']=df5['Humidity3pm'].fillna(df5['Humidity3pm'].mean())

In [ ]: df5['Pressure9am']=df5['Pressure9am'].fillna(df5['Pressure9am'].mean())
```

```
In [ ]: df5['Pressure3pm']=df5['Pressure3pm'].fillna(df5['Pressure3pm'].mean())
```

```
In [ ]: df5['Cloud9am']=df5['Cloud9am'].fillna(df5['Cloud9am'].mean())
```

```
In [ ]: df5['Cloud3pm']=df5['Cloud3pm'].fillna(df5['Cloud3pm'].mean())
```

```
In [ ]: df5['Temp9am']=df5['Temp9am'].fillna(df5['Temp9am'].mean())
```

```
In [ ]: df5['Temp3pm']=df5['Temp3pm'].fillna(df5['Temp3pm'].mean())
```

```
In [ ]: df5['Temp3pm']=df5['Temp3pm'].fillna(df5['Temp3pm'].mean())
```

```
In [ ]: df5['RainToday'].unique()
```

```
Out[ ]: array(['No', 'Yes', nan], dtype=object)
```

```
In [ ]: df5['RainToday']=df5['RainToday'].fillna(np.random.choice(['No', 'Yes']))
```

```
In [ ]: df5.isnull().sum()
```

```
Out[ ]: 0
         Date 0
         Location 0
         MinTemp 0
         MaxTemp 0
         Rainfall 0
         Evaporation 0
         Sunshine 0
         WindGustDir 0
         WindGustSpeed 0
         WindDir9am 0
         WindDir3pm 0
         WindSpeed9am 0
         WindSpeed3pm 0
         Humidity9am 0
         Humidity3pm 0
         Pressure9am 0
         Pressure3pm 0
         Cloud9am 0
         Cloud3pm 0
         Temp9am 0
         Temp3pm 0
         RainToday 0
         RISK_MM 1
         RainTomorrow 1
```

**dtype:** int64

```
In [ ]: df5.duplicated().sum()
```

```
Out[ ]: np.int64(0)
```

```
In [ ]: df5.describe()
```

Out[ ]:

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	W
<b>count</b>	94773.000000	94773.000000	94773.000000	94773.000000	94773.000000	
<b>mean</b>	12.086363	22.760775	2.535371	5.228030	7.343673	
<b>std</b>	6.437633	6.852767	9.110664	2.903462	2.653386	
<b>min</b>	-8.500000	-4.800000	0.000000	0.000000	0.000000	
<b>25%</b>	7.500000	17.800000	0.000000	4.000000	7.343673	
<b>50%</b>	12.100000	22.600000	0.000000	5.228030	7.343673	
<b>75%</b>	17.000000	27.600000	1.000000	5.228030	8.000000	
<b>max</b>	33.900000	47.300000	371.000000	145.000000	14.500000	

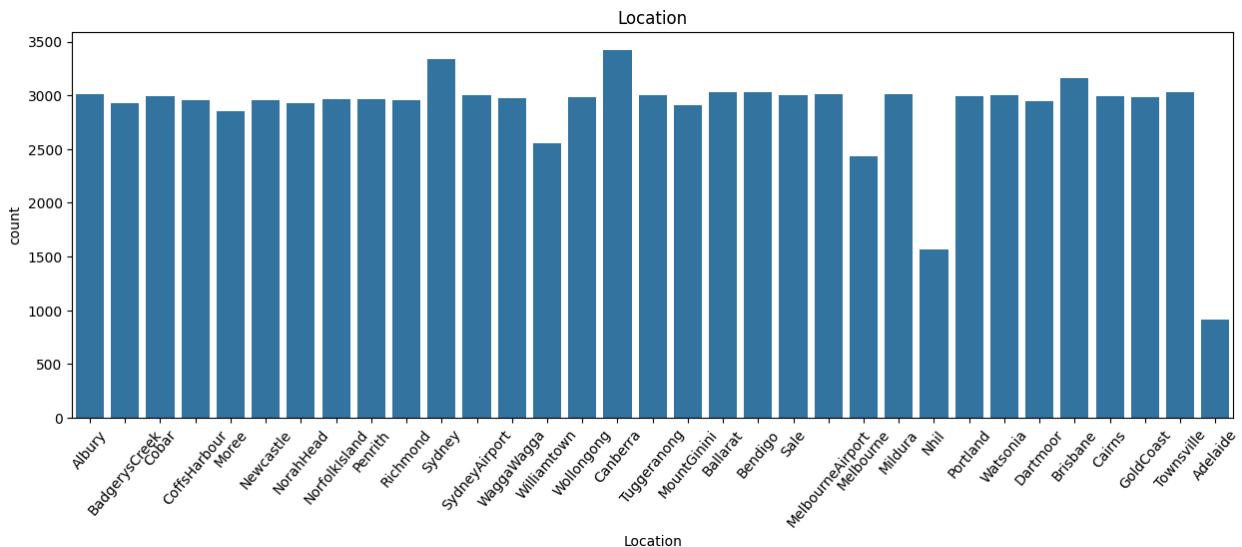
In [ ]: categorical5=df5.select\_dtypes(include='object')

In [ ]: 

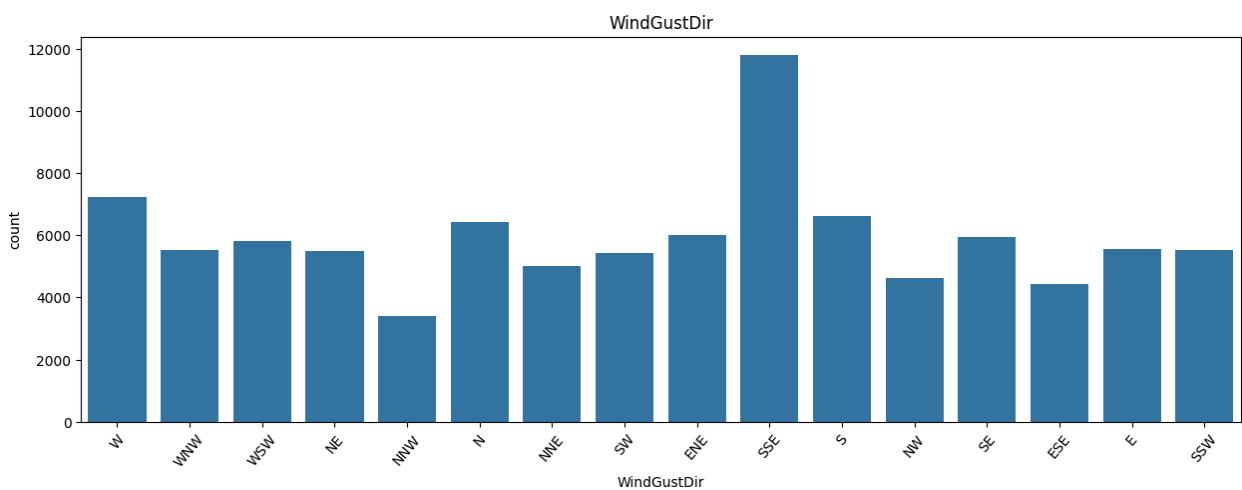
```
for i in categorical5.columns:
    print(df5[i].describe())
    print('*****')
```

```
count          94773
unique         3436
top      2010-12-05
freq            32
Name: Date, dtype: object
*****
count          94773
unique         33
top      Canberra
freq        3418
Name: Location, dtype: object
*****
count          94773
unique         16
top       SSE
freq      11791
Name: WindGustDir, dtype: object
*****
count          94773
unique         16
top       NNE
freq      12703
Name: WindDir9am, dtype: object
*****
count          94773
unique         16
top       NE
freq      9527
Name: WindDir3pm, dtype: object
*****
count          94773
unique         2
top       No
freq      73278
Name: RainToday, dtype: object
*****
count          94772
unique         2
top       No
freq      72957
Name: RainTomorrow, dtype: object
*****
```

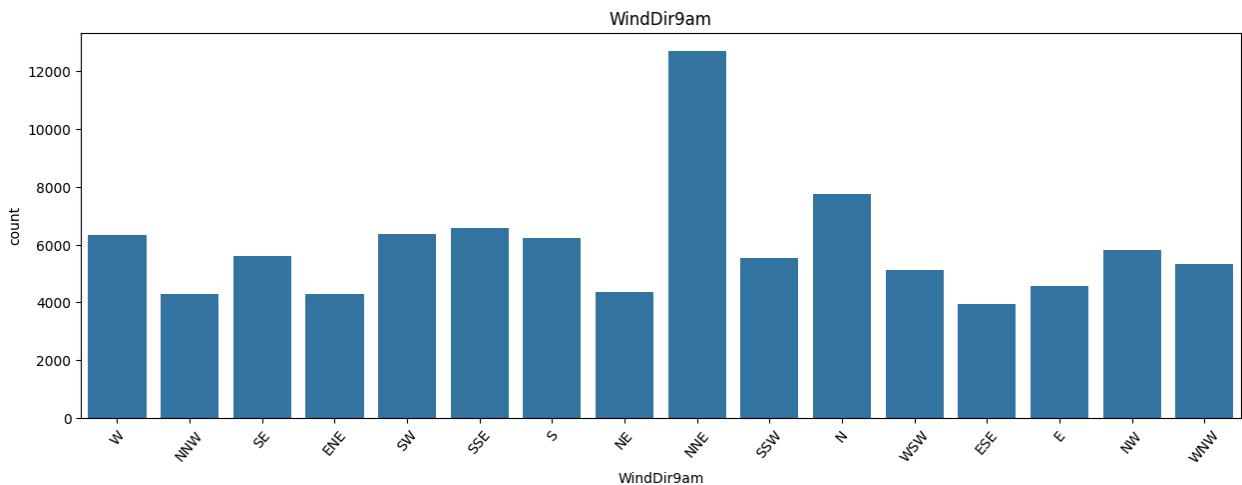
```
In [ ]: for i in categorical5:
    if i != 'Date':
        plt.figure(figsize=(15,5))
        sns.countplot(x=df5[i])
        plt.title(i)
        plt.xticks(rotation=50)
        plt.show()
        print('*****')
```



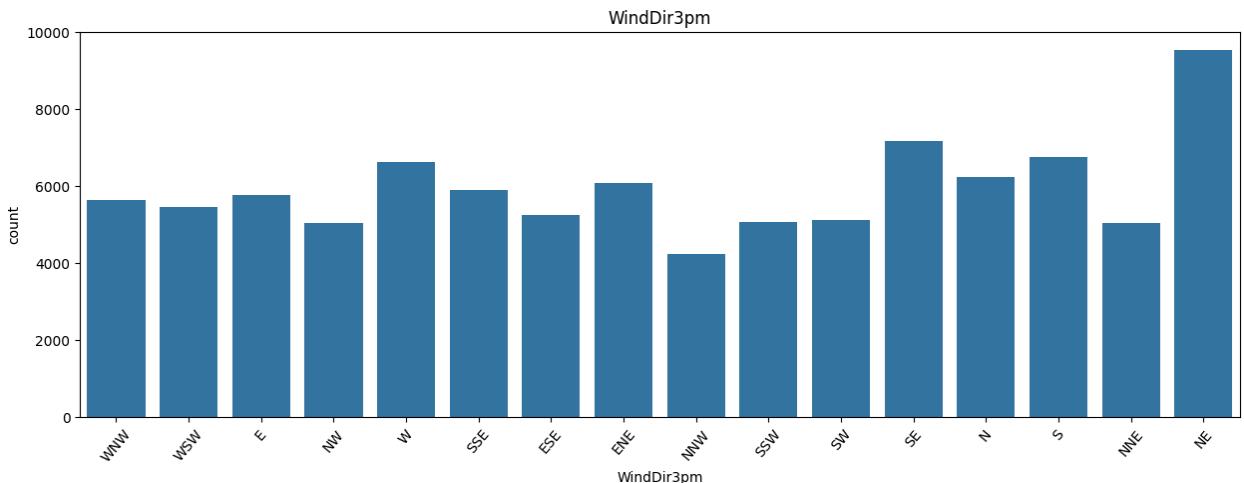
\*\*\*\*\*



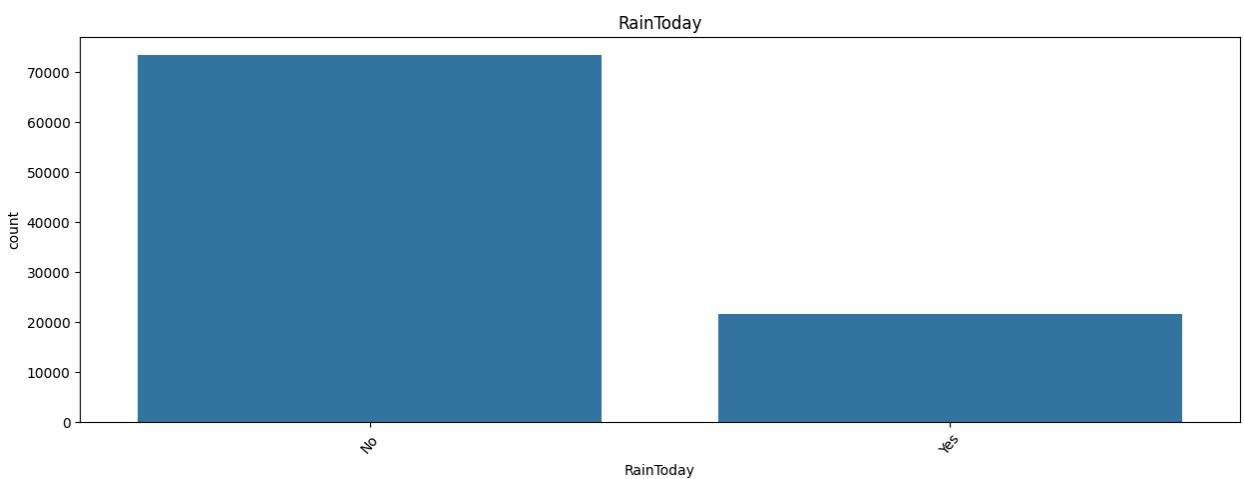
\*\*\*\*\*



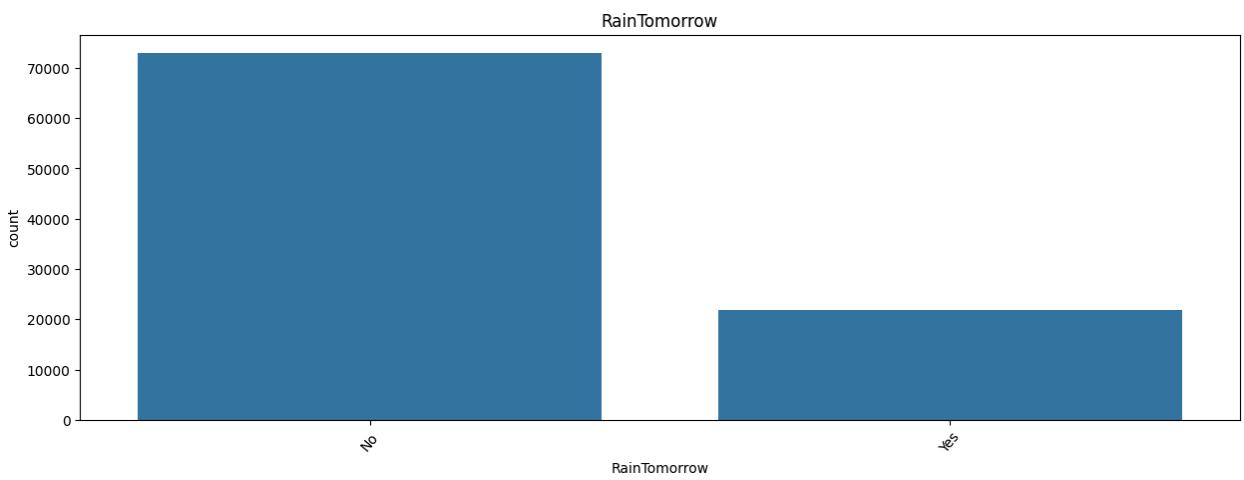
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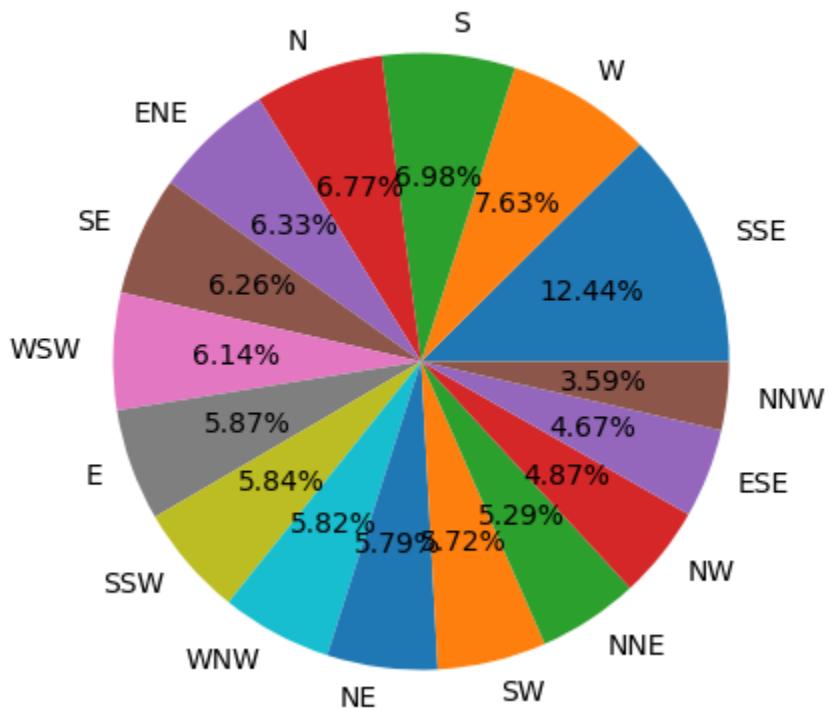


\*\*\*\*\*

```
In [ ]: for i in categorical5:
    if i!='Date' and i!='Location':
        plt.figure(figsize=(15,5))
        plt.pie(df5[i].value_counts(), labels=df5[i].value_counts().index, autopct='%1.1f%%')
        plt.title(i)
        plt.xticks(rotation=50)
```

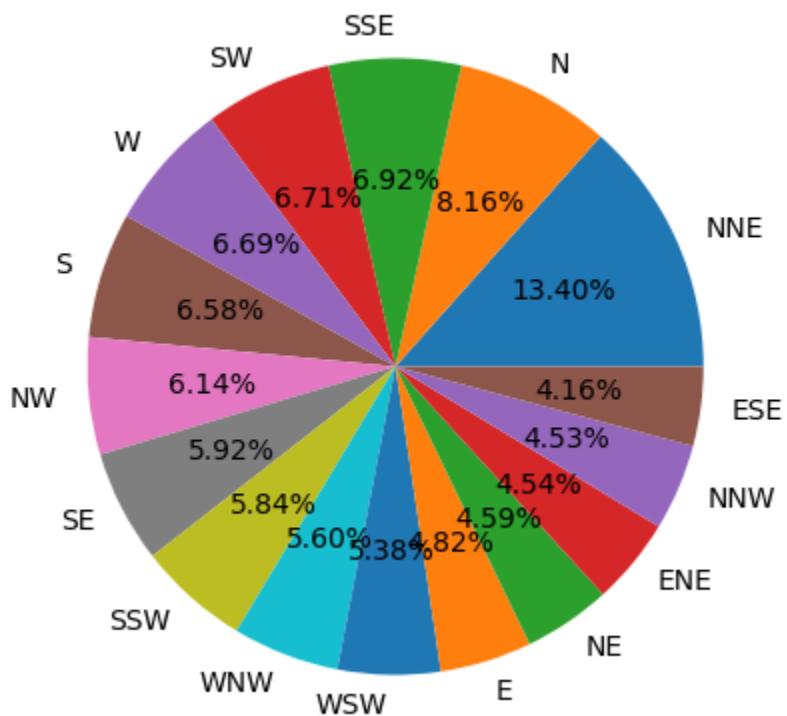
```
plt.show()  
print('*****')
```

WindGustDir



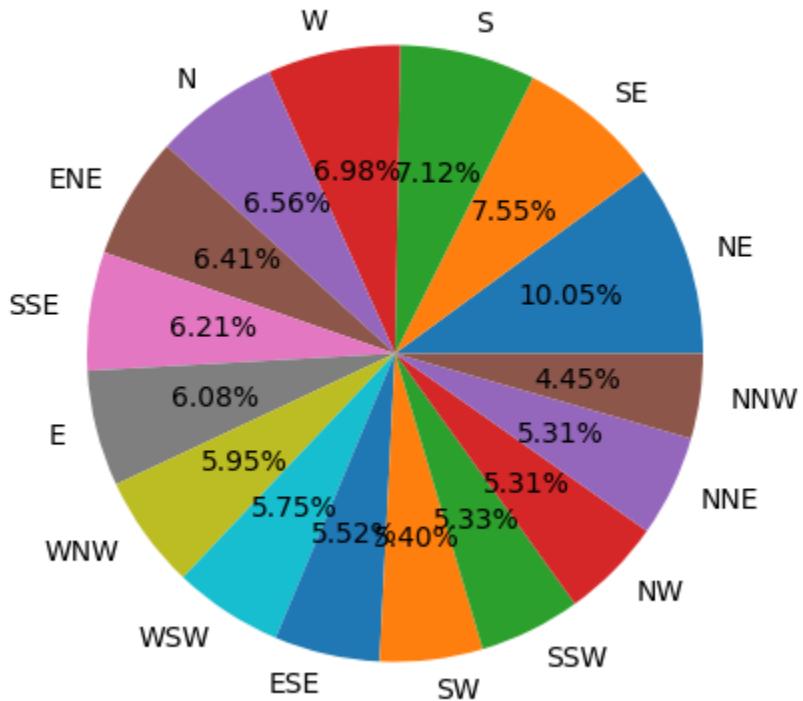
\*\*\*\*\*

WindDir9am



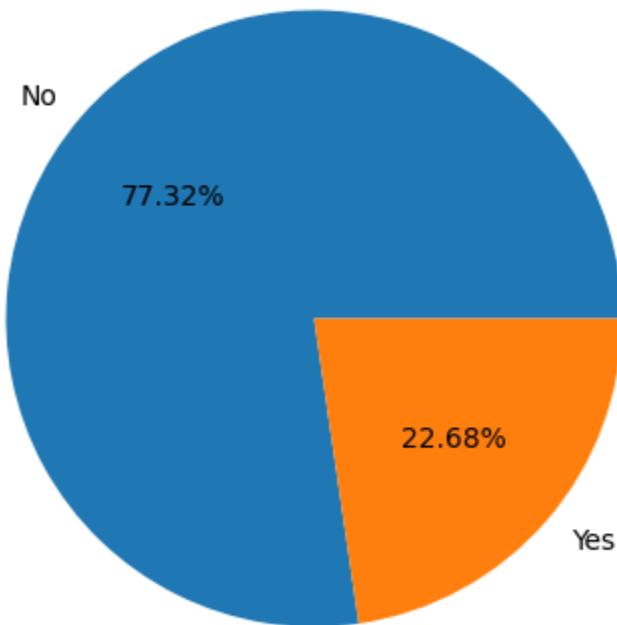
\*\*\*\*\*

WindDir3pm



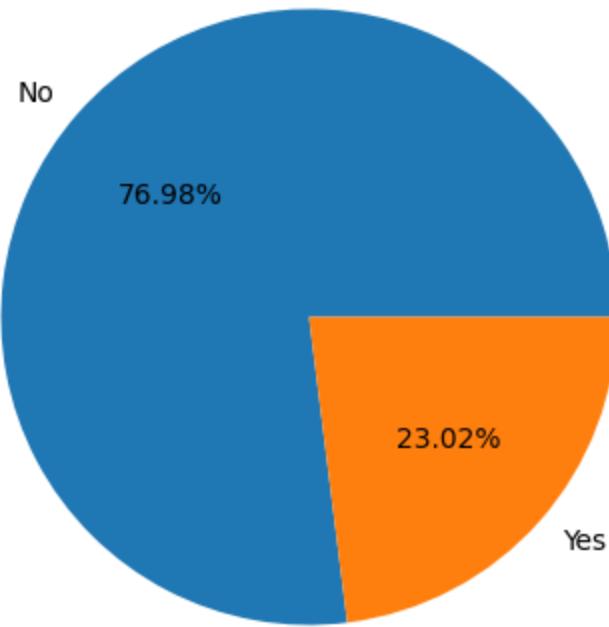
\*\*\*\*\*

RainToday



\*\*\*\*\*

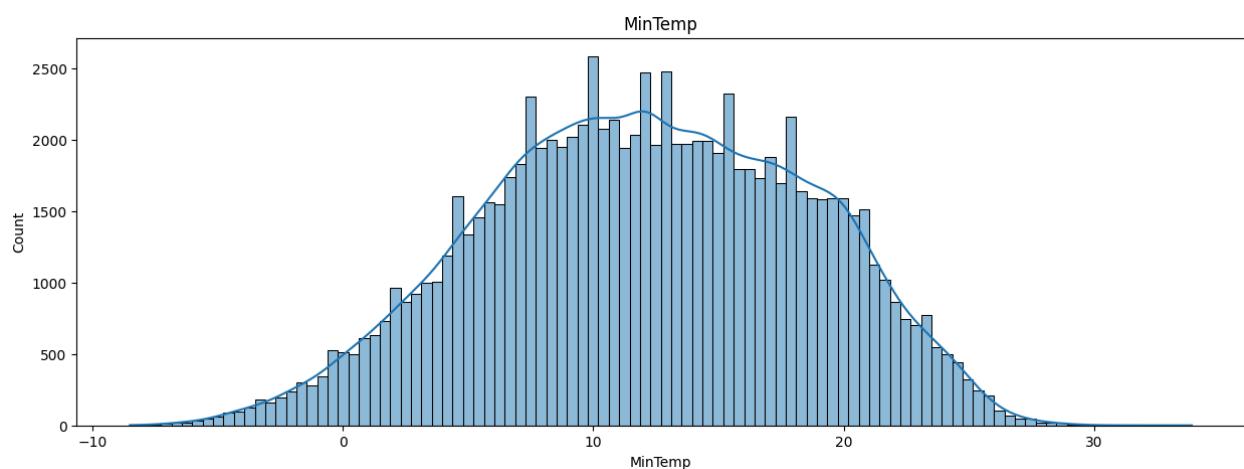
## RainTomorrow

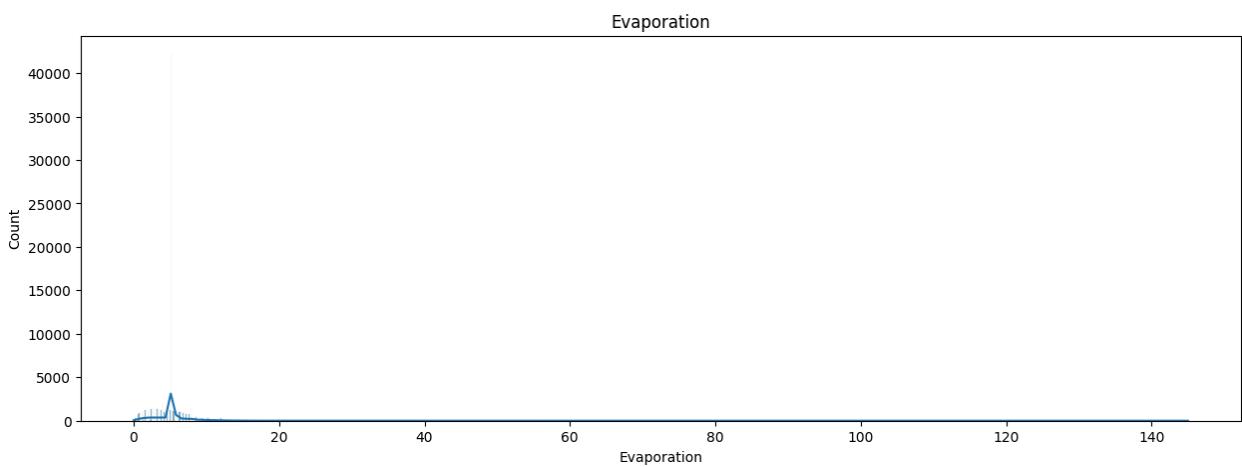
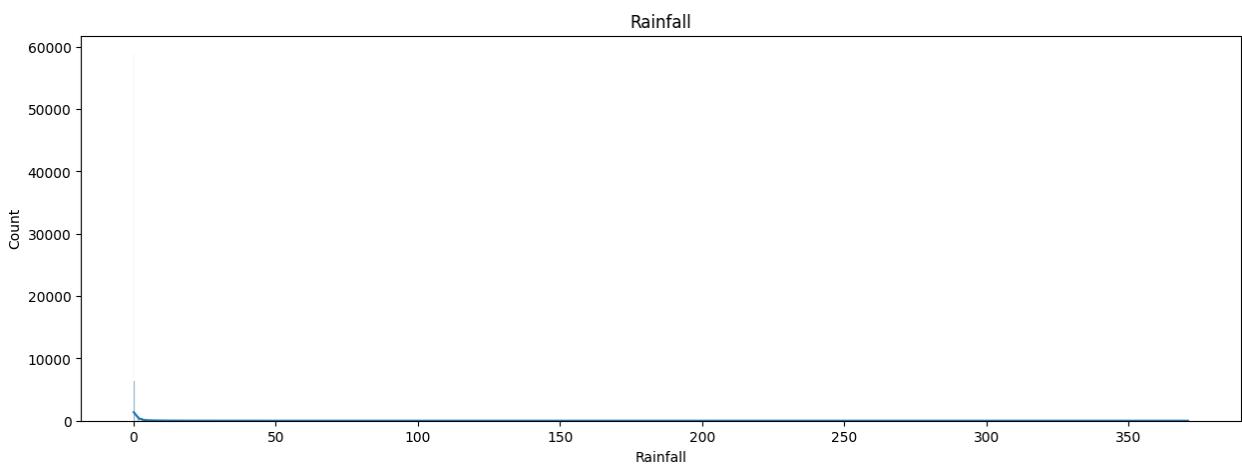
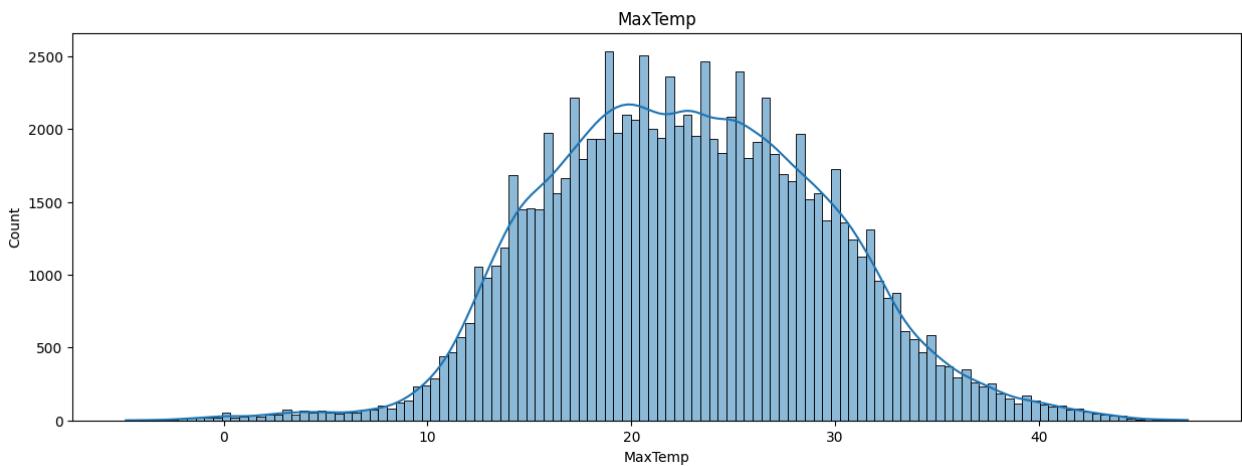


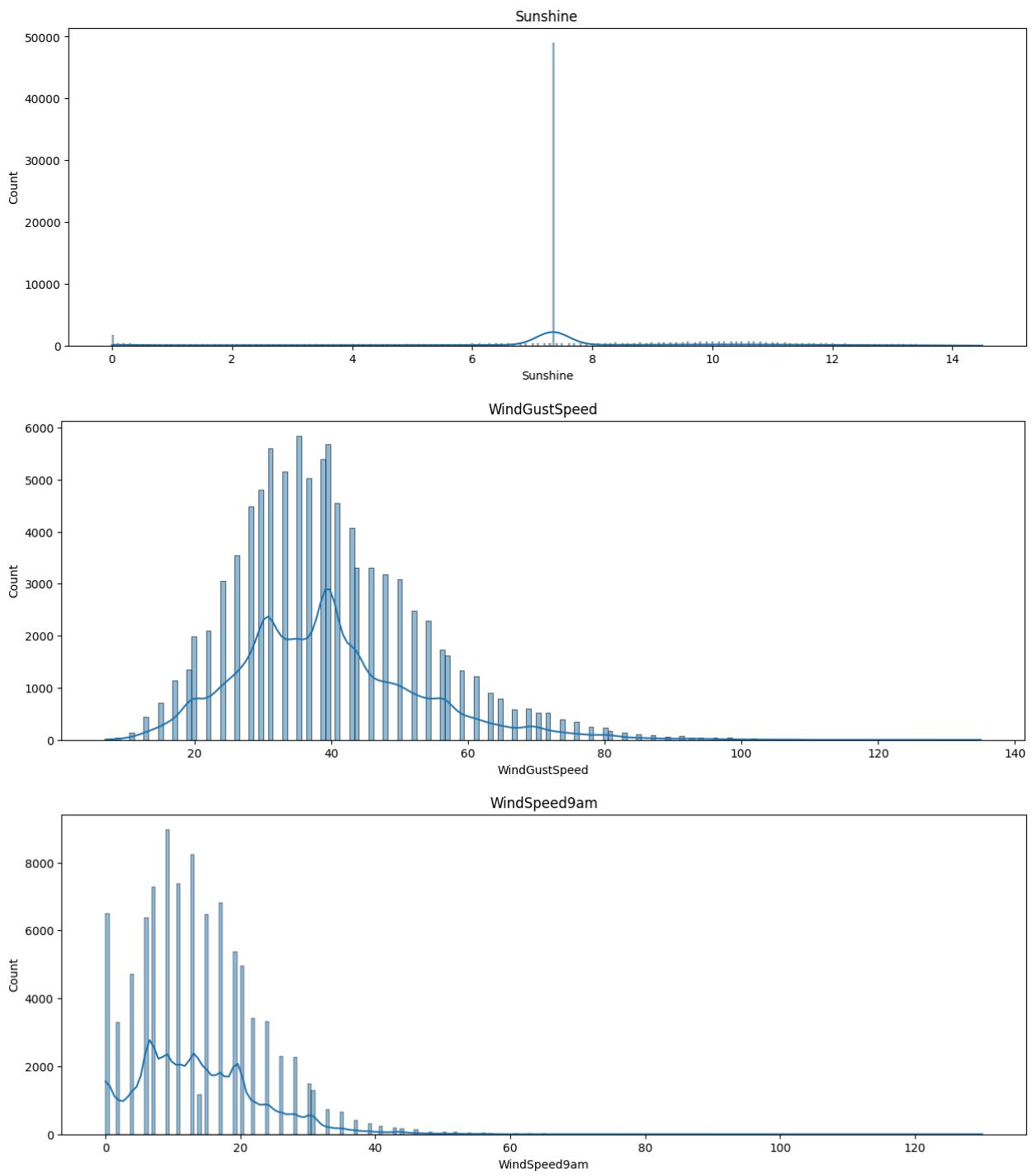
\*\*\*\*\*

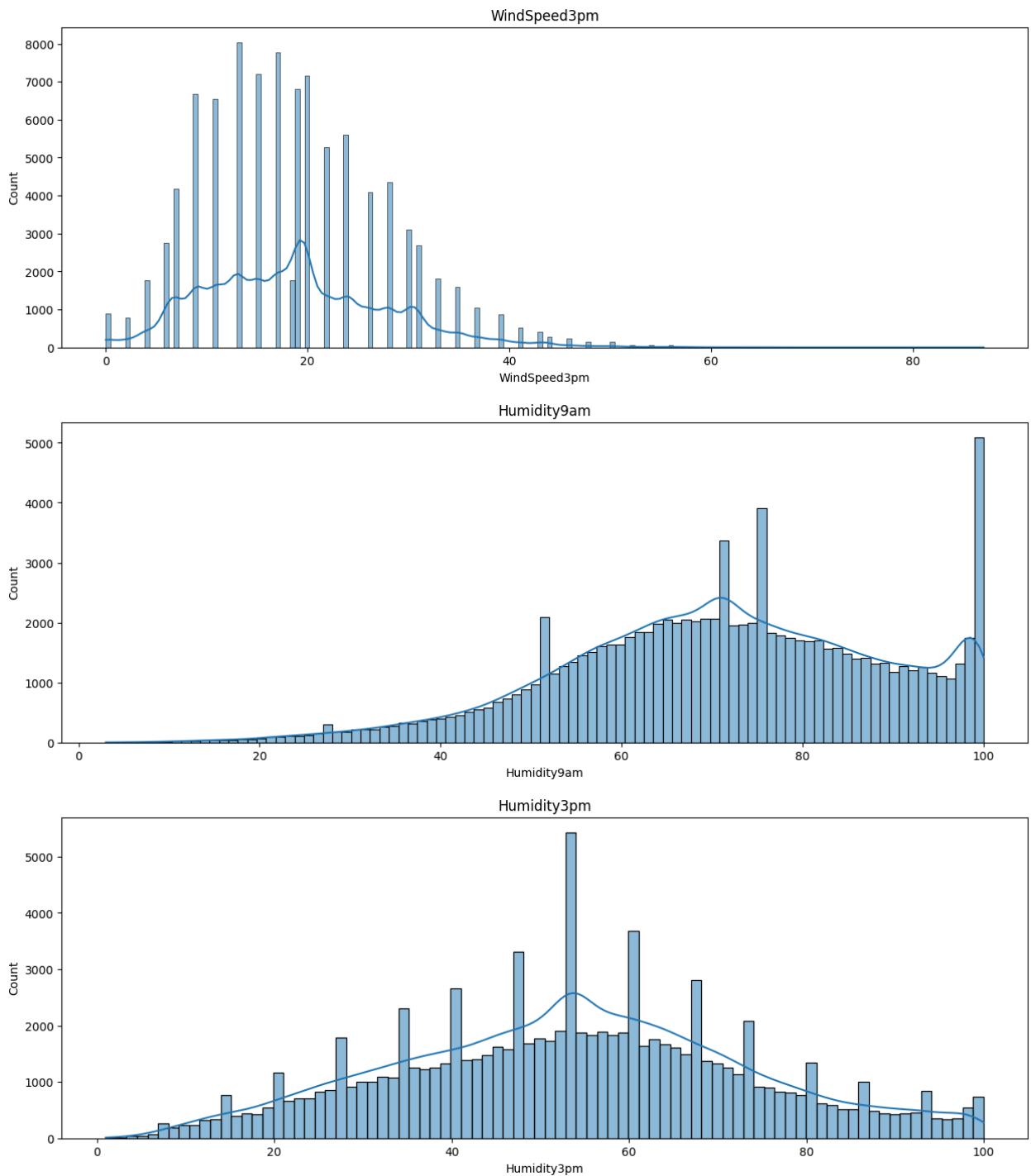
```
In [ ]: Numerical5=df5.select_dtypes(exclude='object')
```

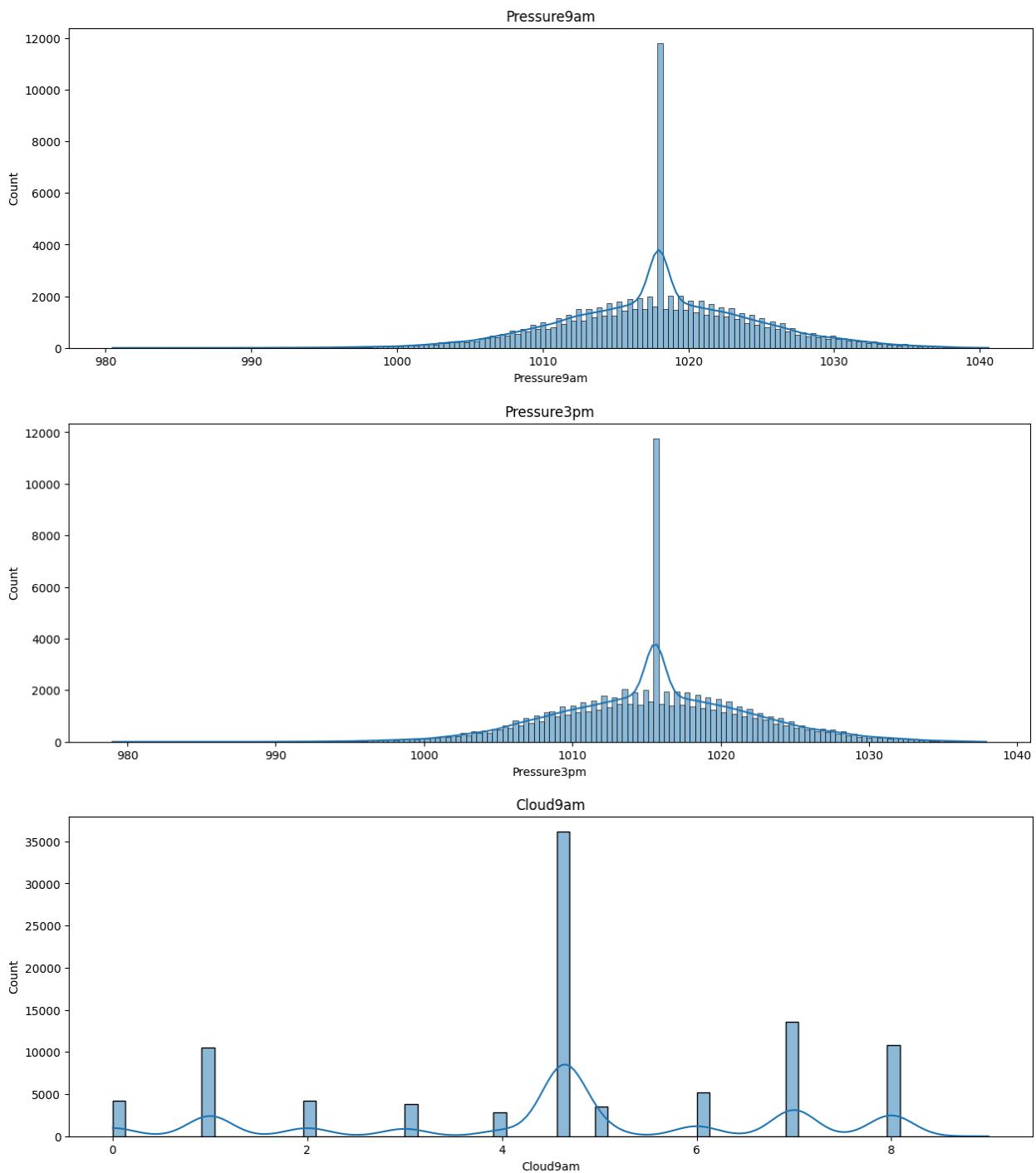
```
In [ ]: for i in Numerical5:  
    plt.figure(figsize=(15,5))  
    sns.histplot(x=df5[i],kde=True)  
    plt.title(i)  
    plt.show()
```

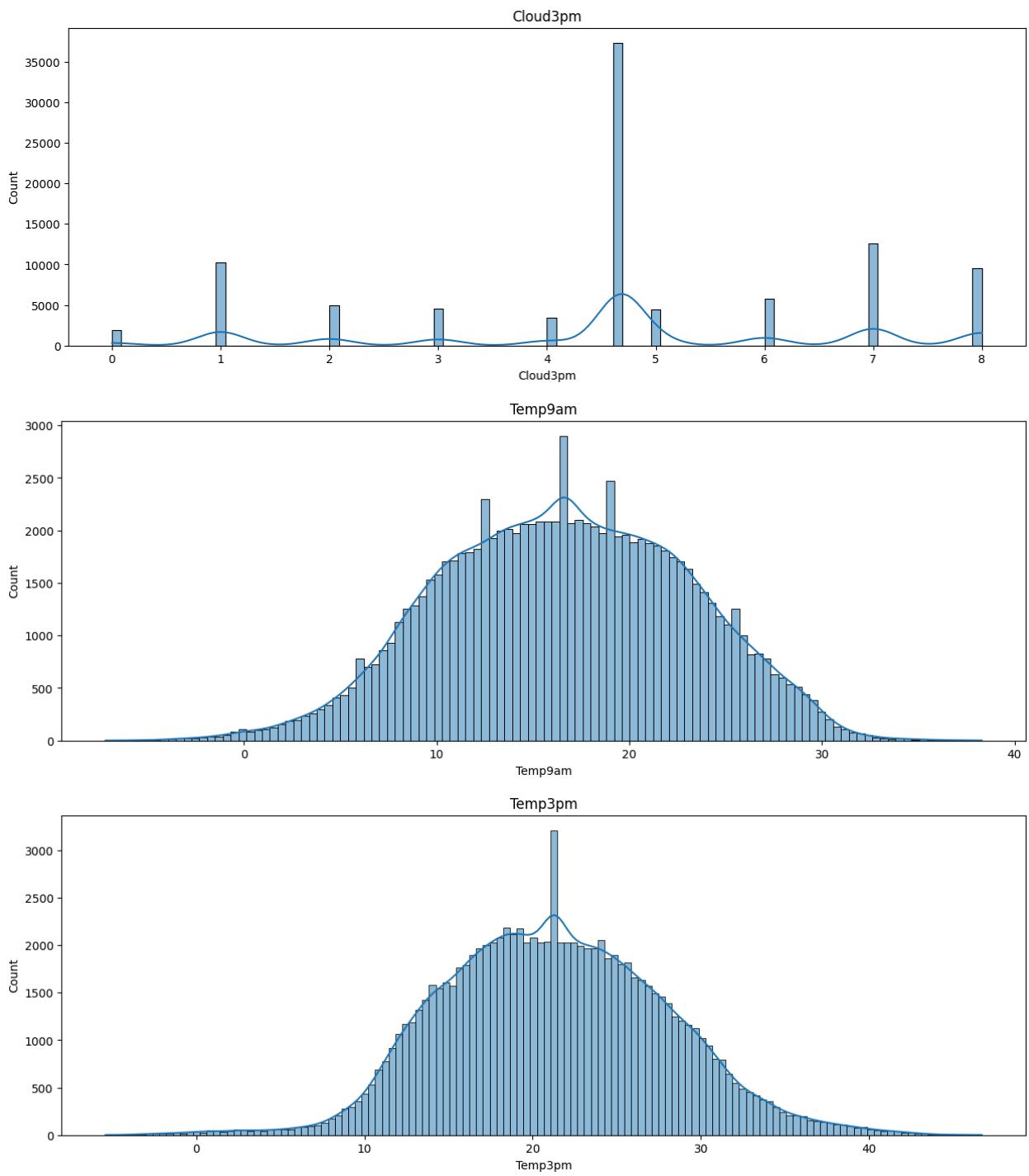


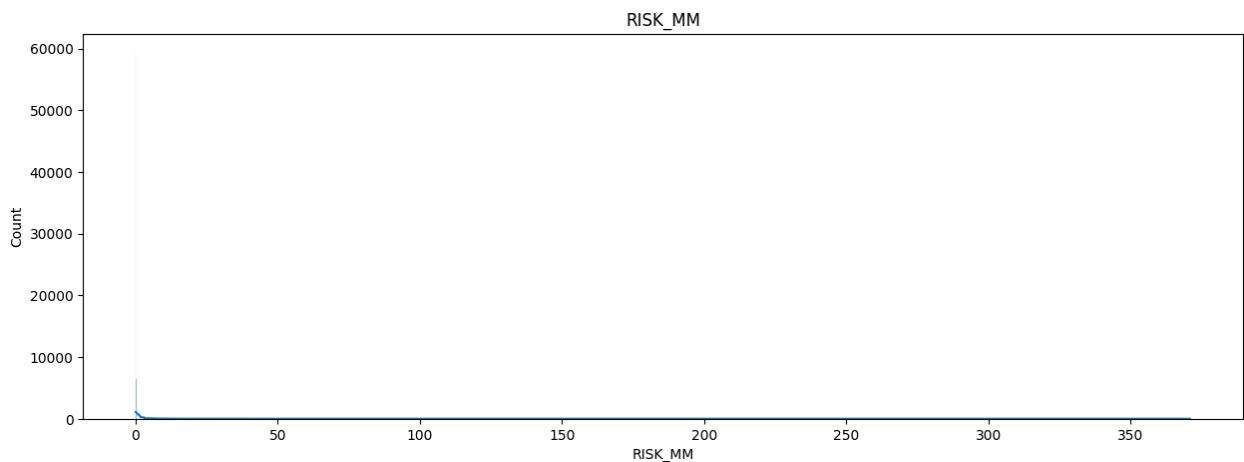




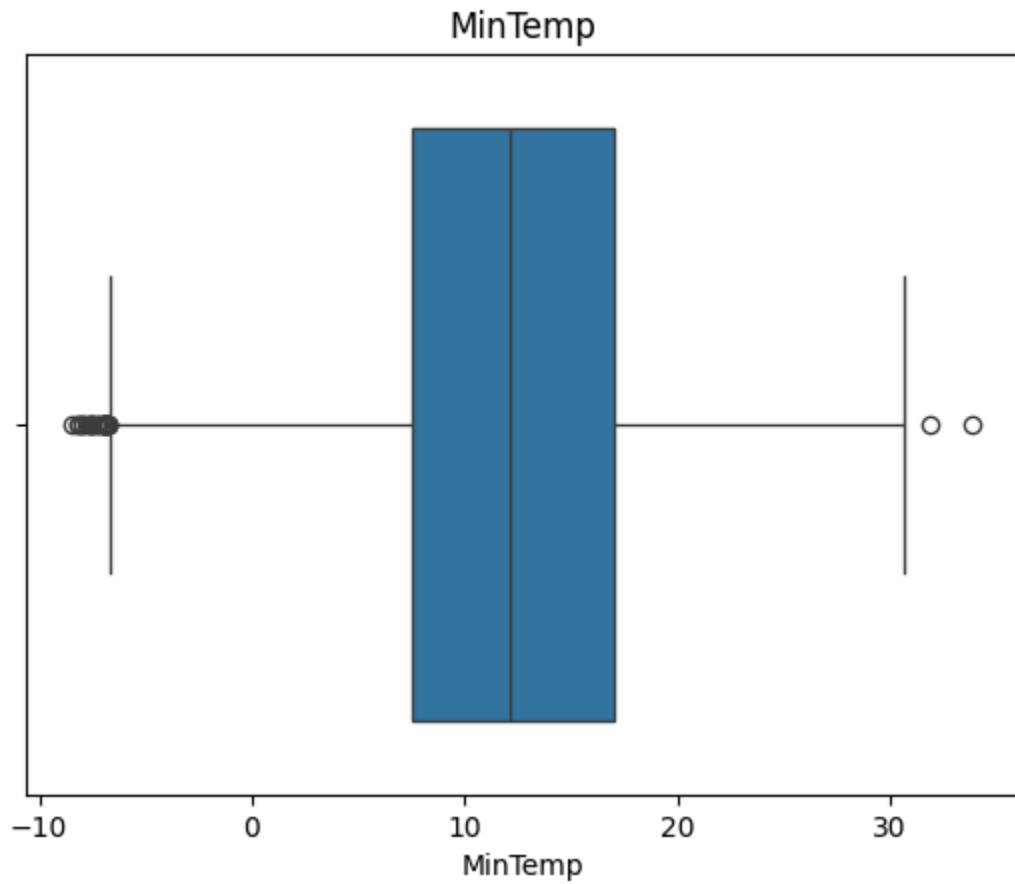




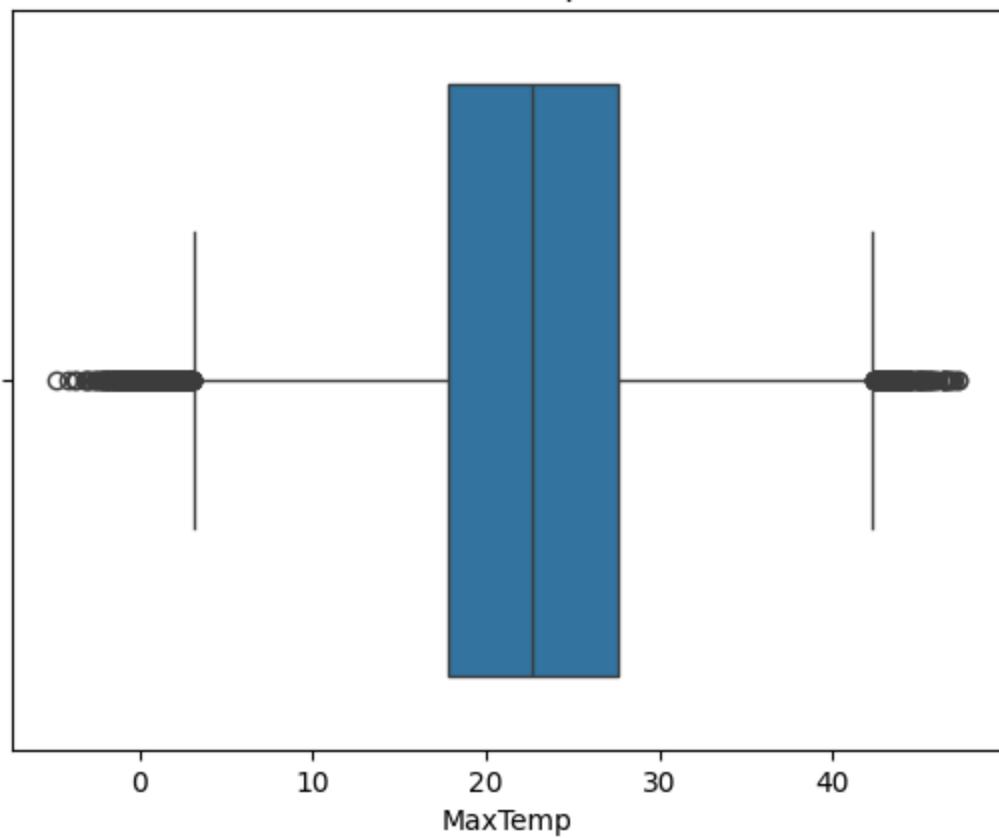




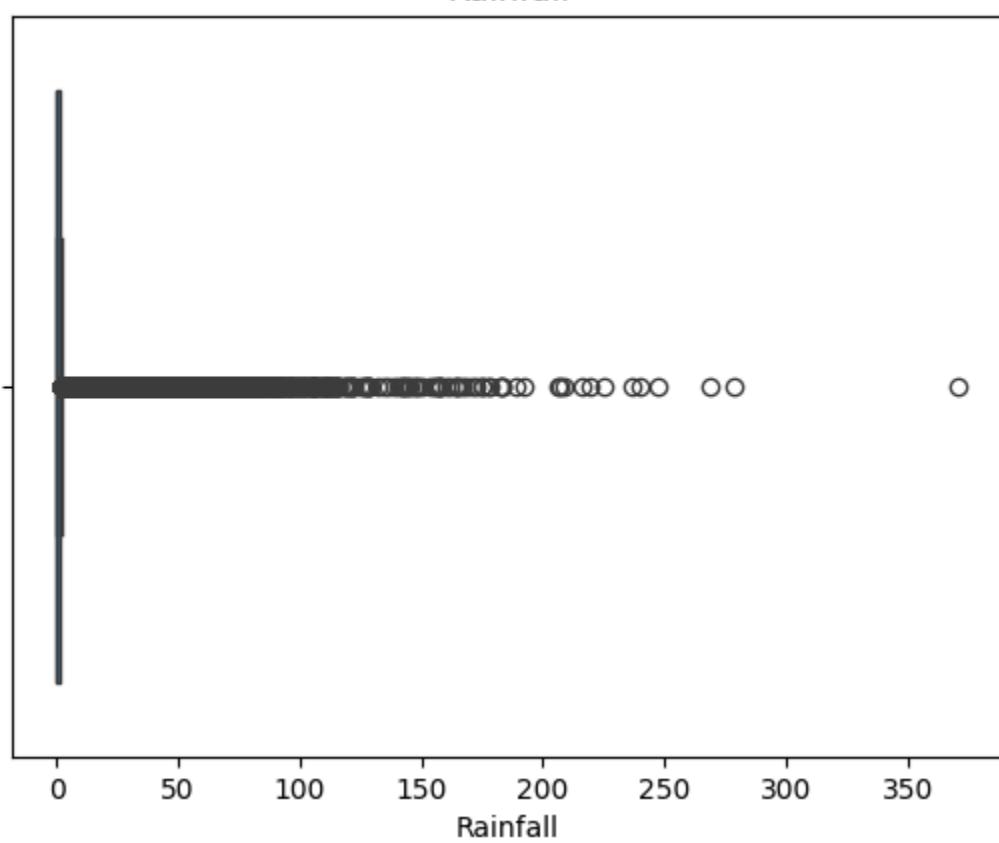
```
In [ ]: for i in Numerical5:  
    sns.boxplot(x=df5[i])  
    plt.title(i)  
    plt.show()
```



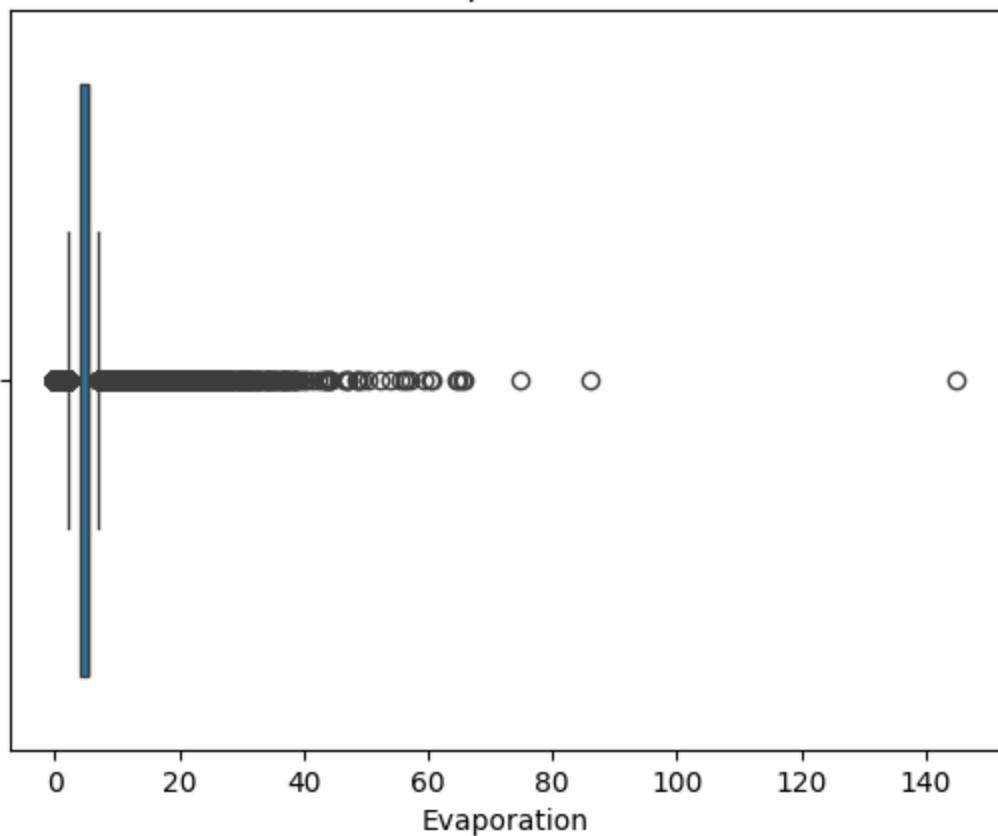
MaxTemp



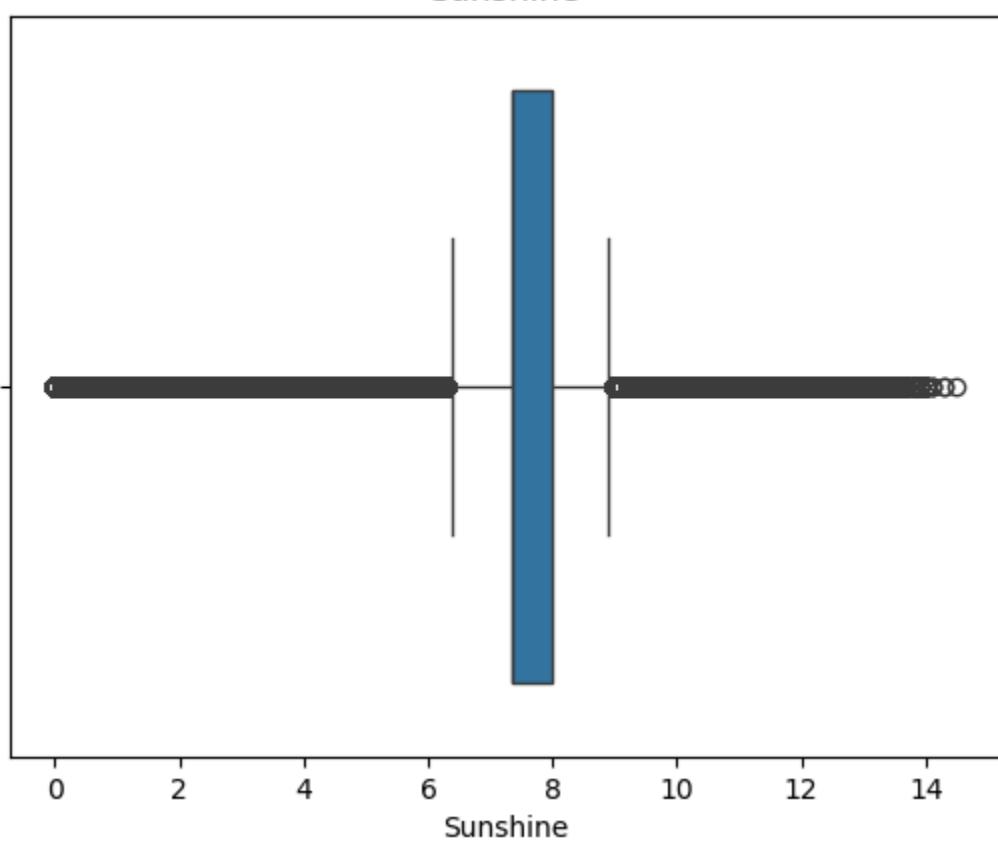
Rainfall



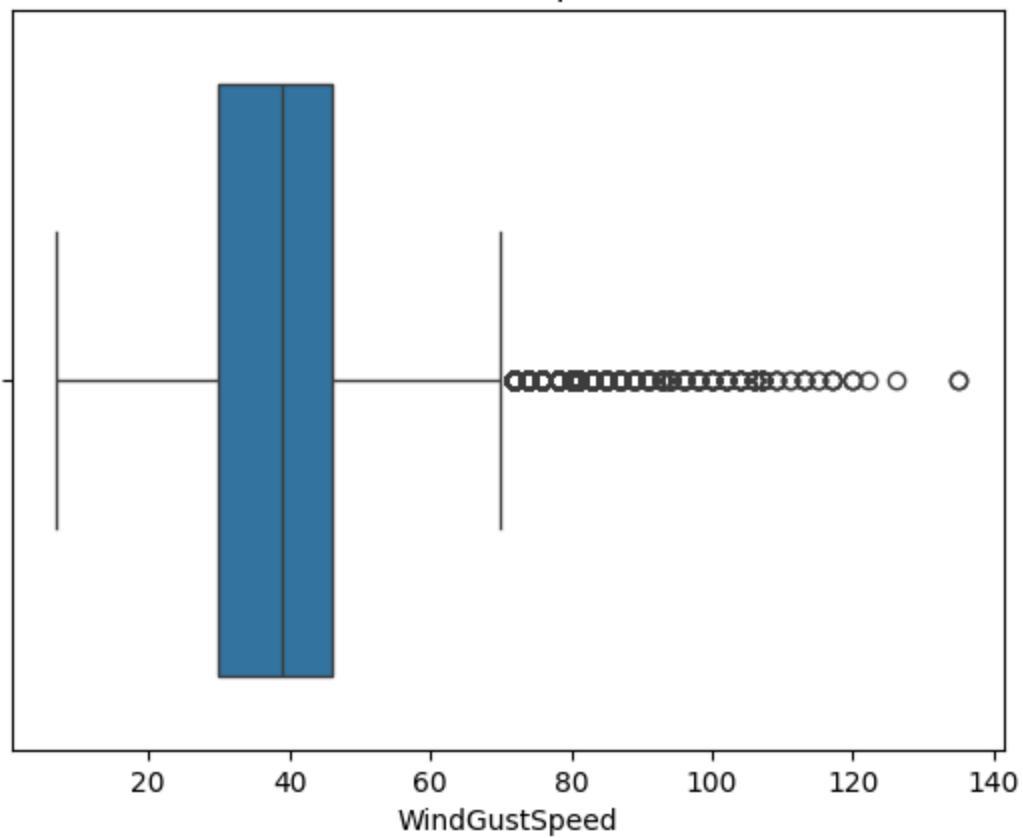
Evaporation



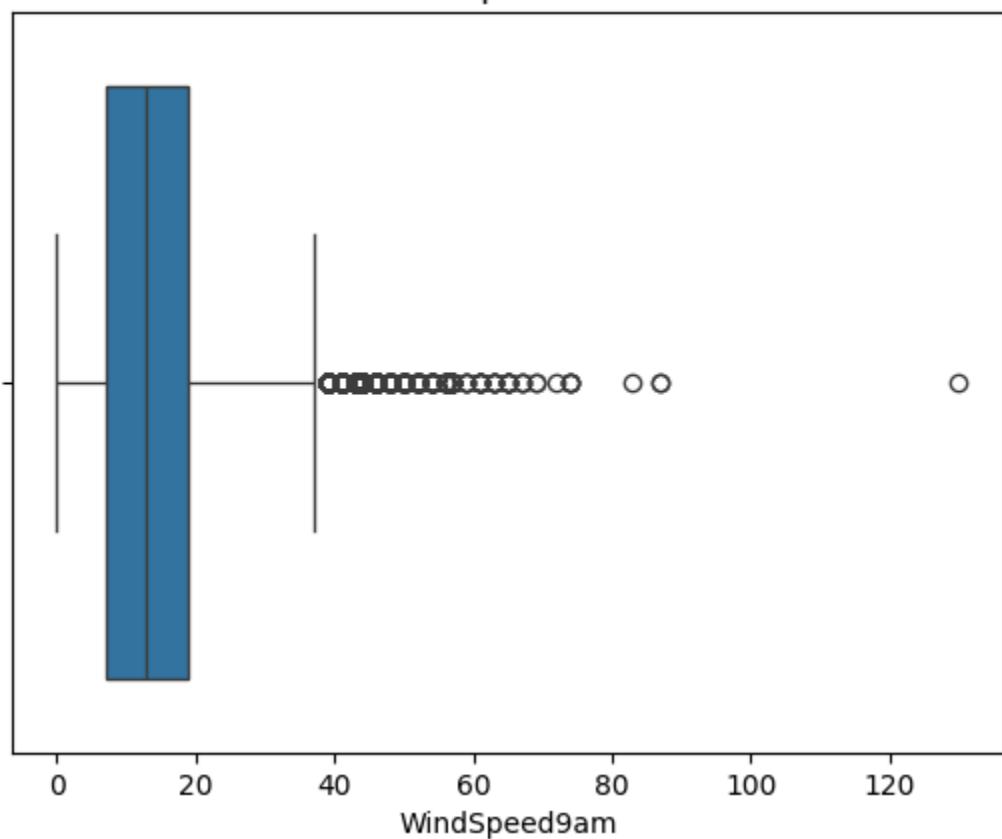
Sunshine



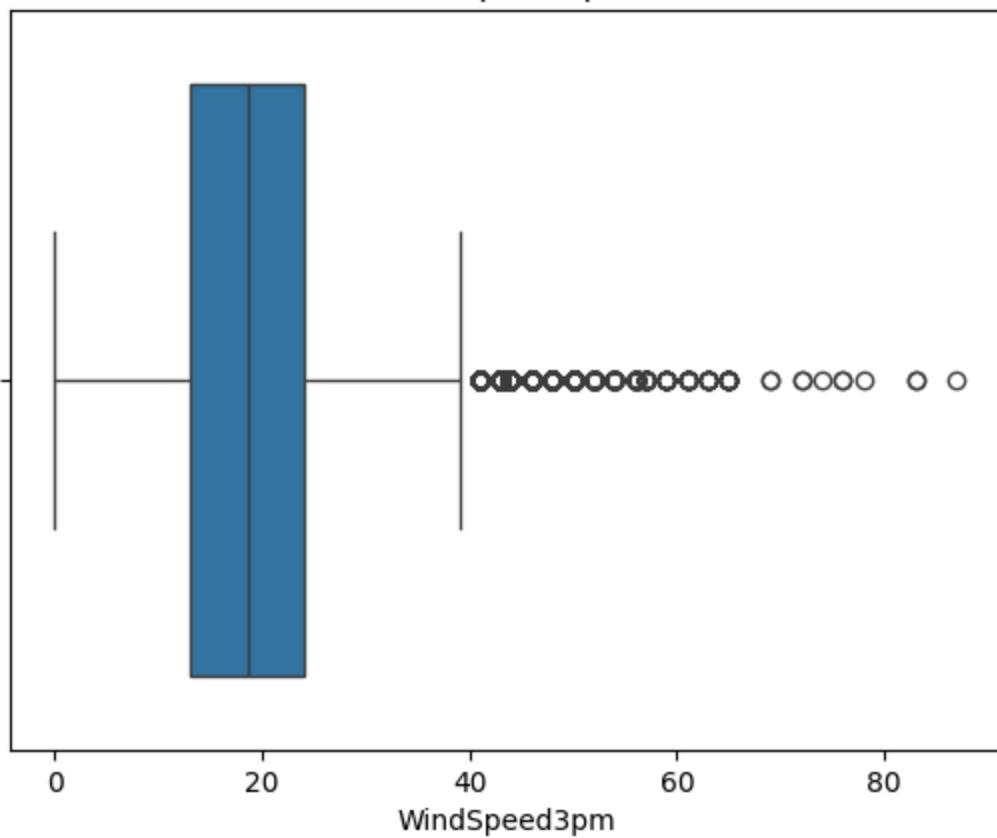
WindGustSpeed



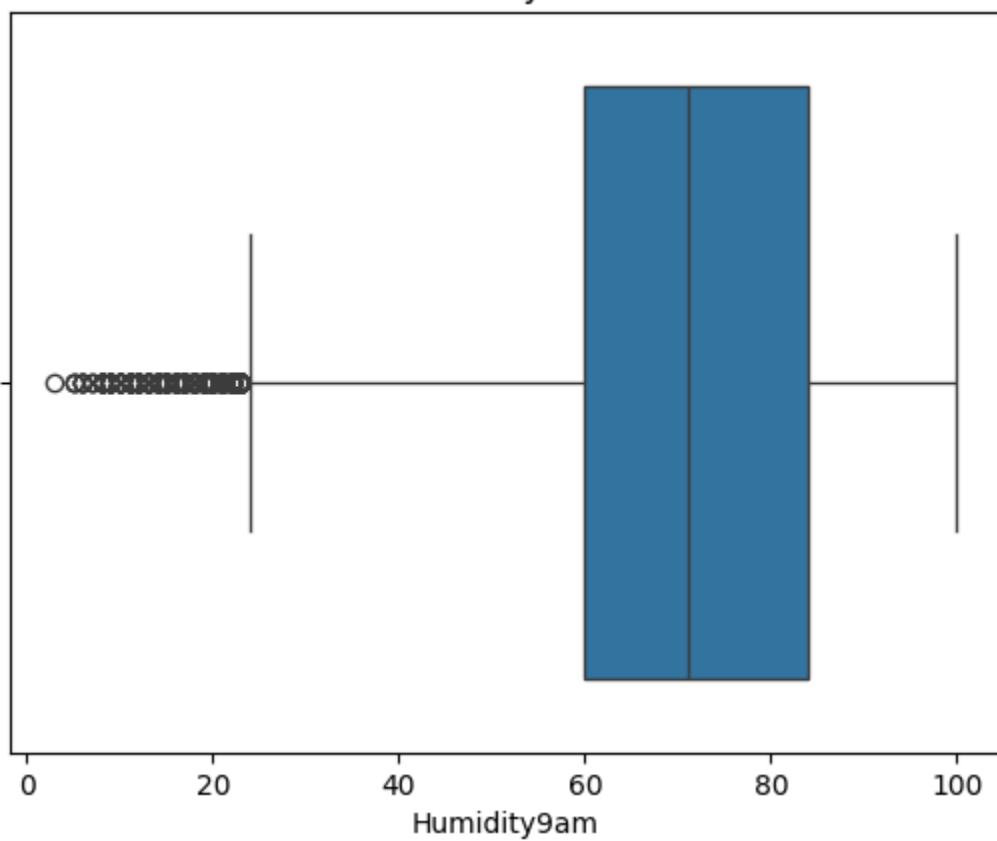
WindSpeed9am



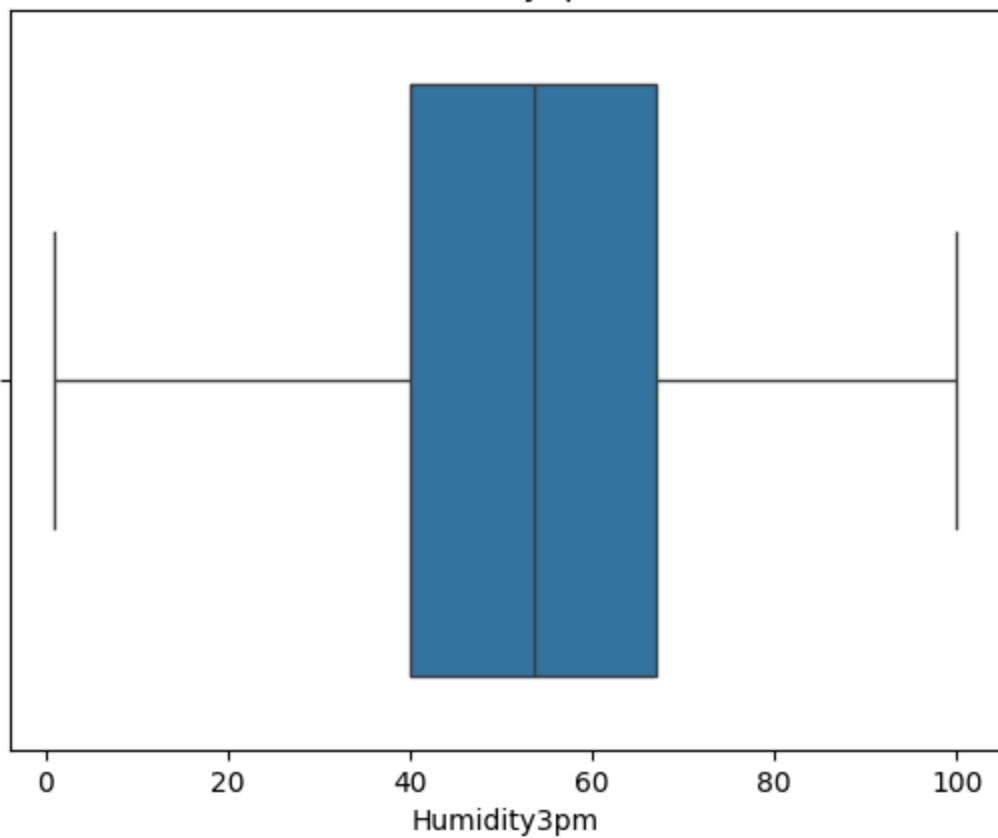
WindSpeed3pm



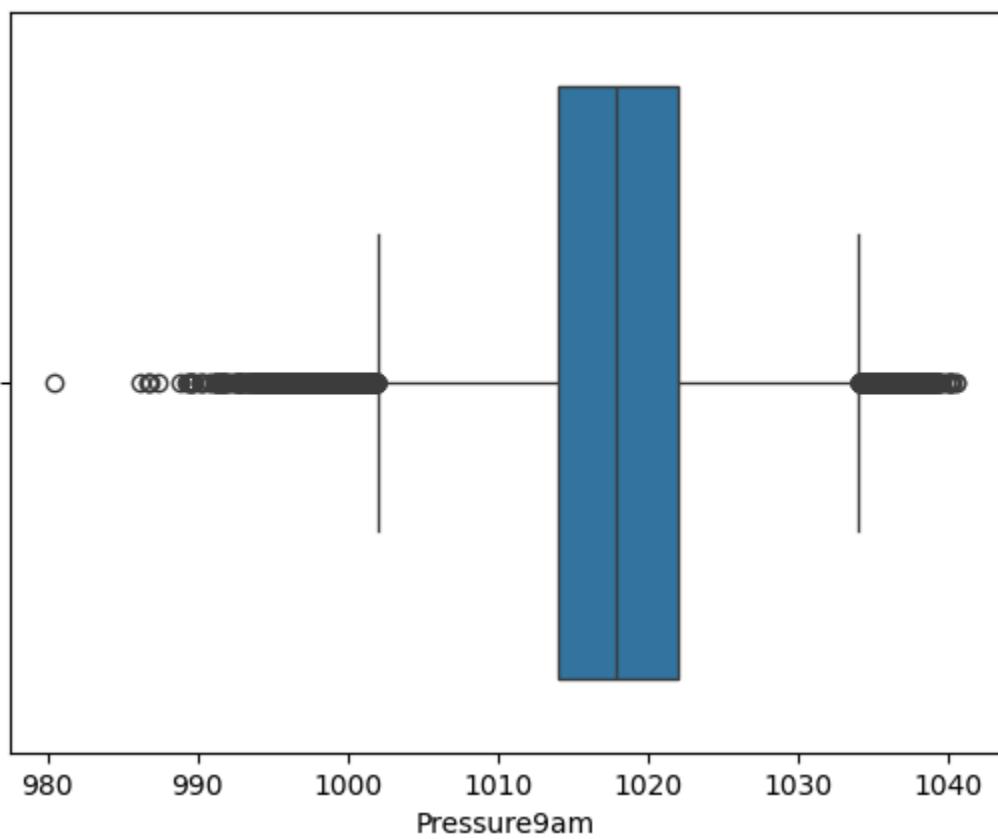
Humidity9am



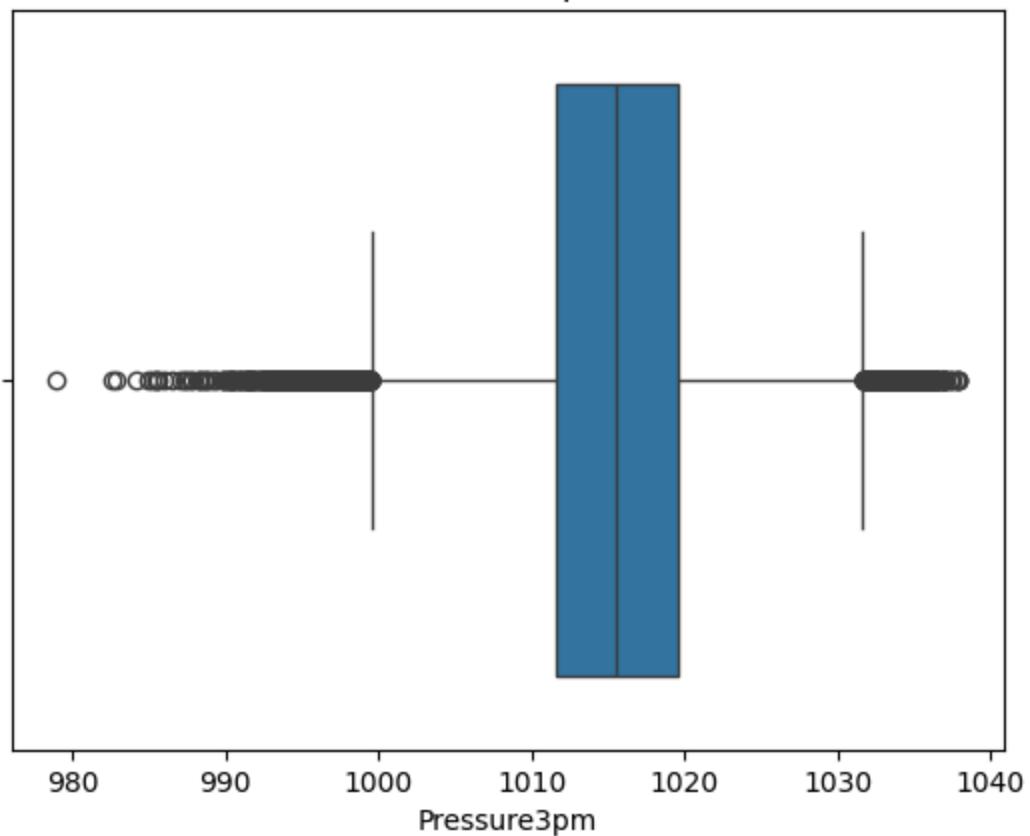
Humidity3pm



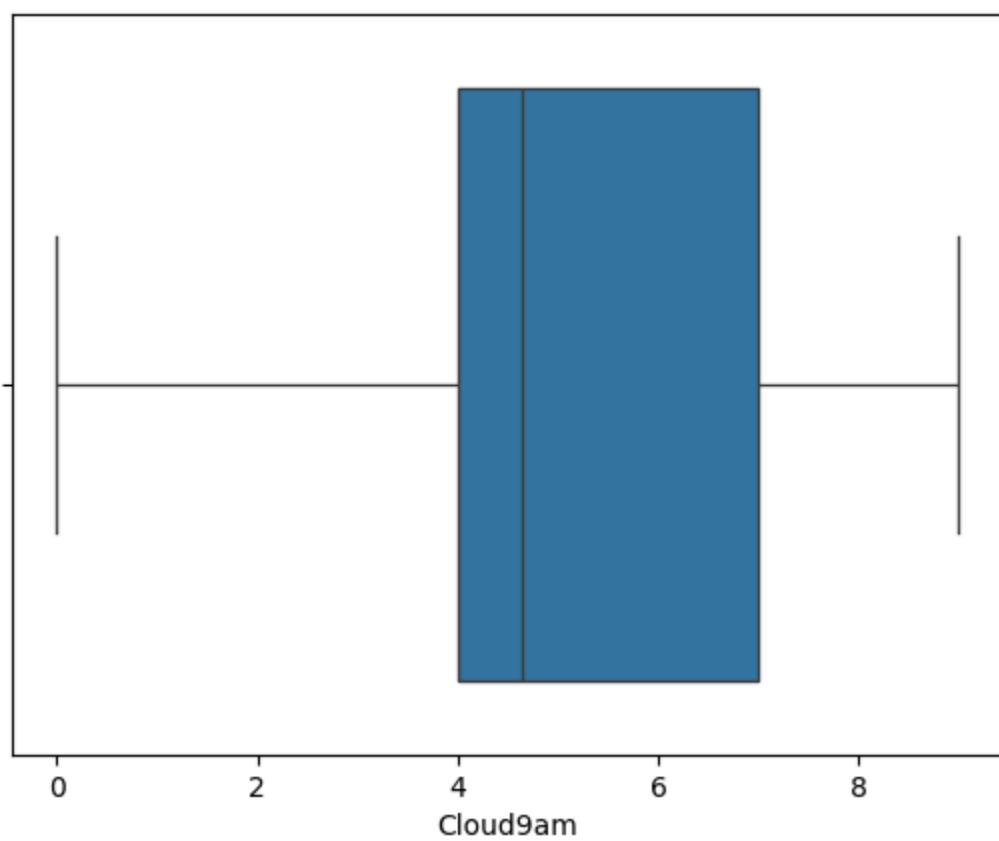
Pressure9am



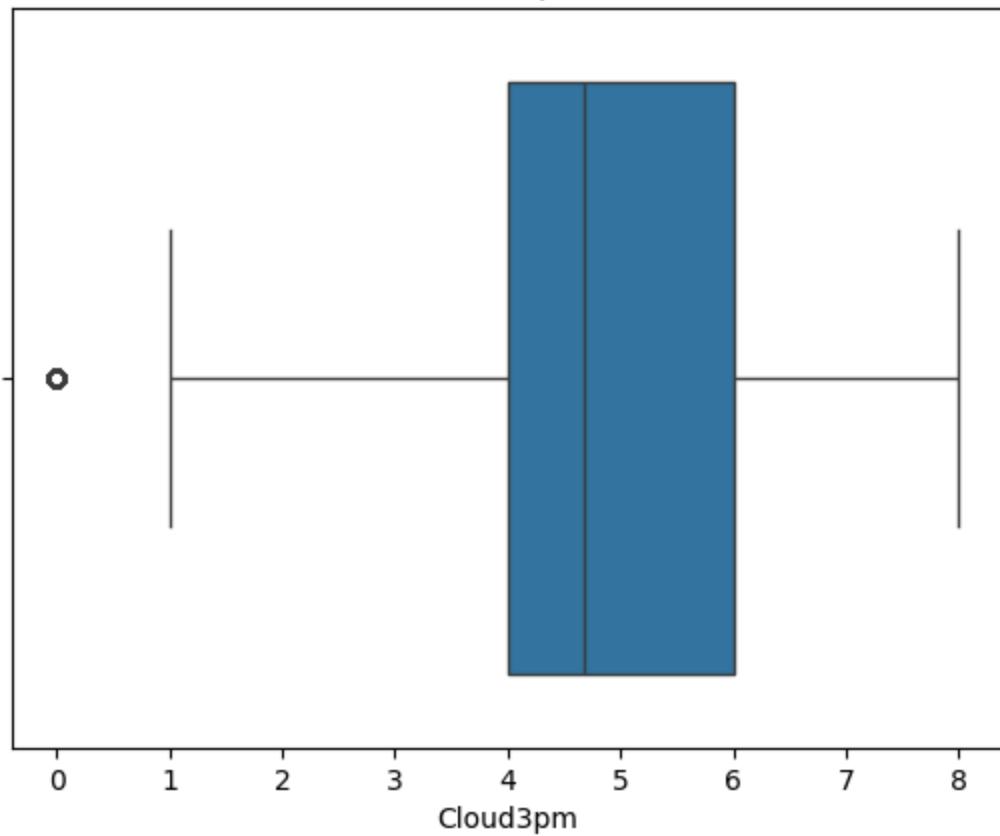
Pressure3pm



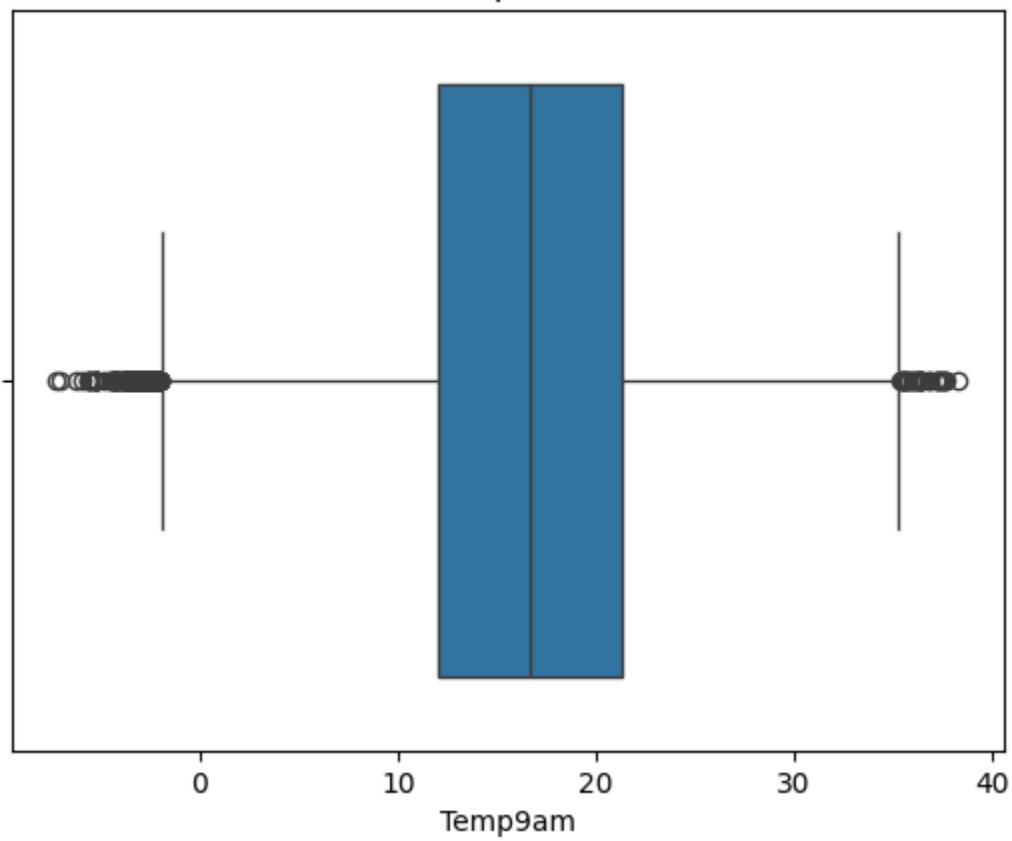
Cloud9am



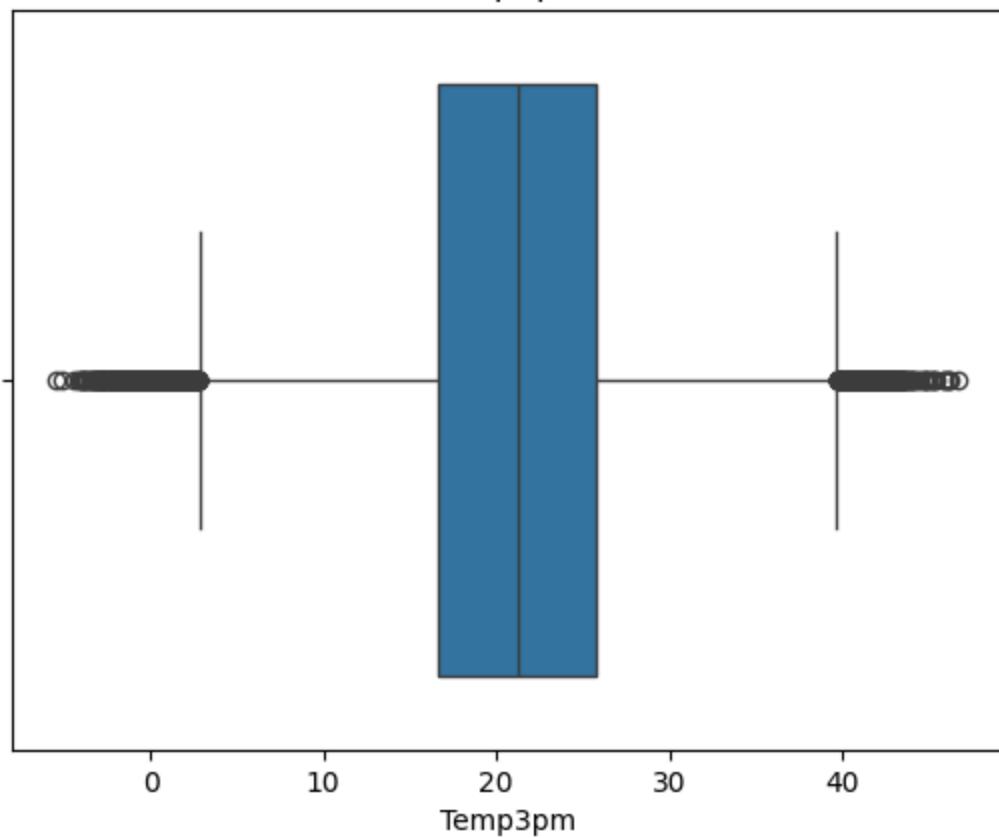
Cloud3pm



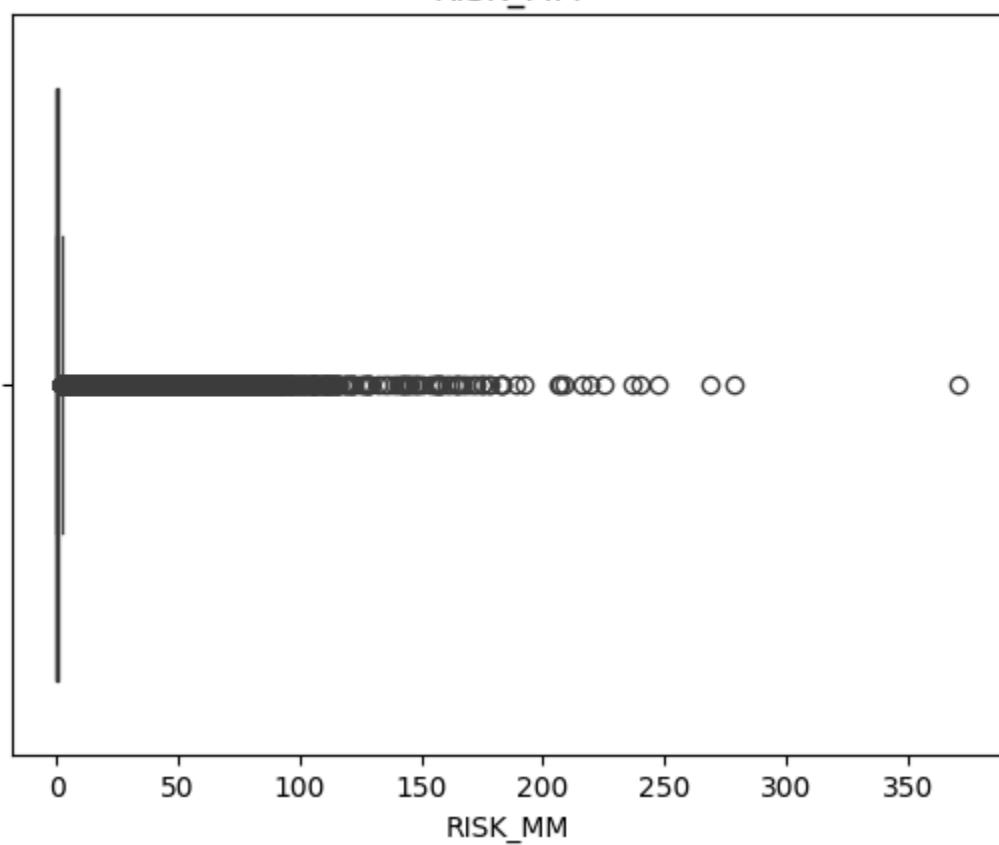
Temp9am



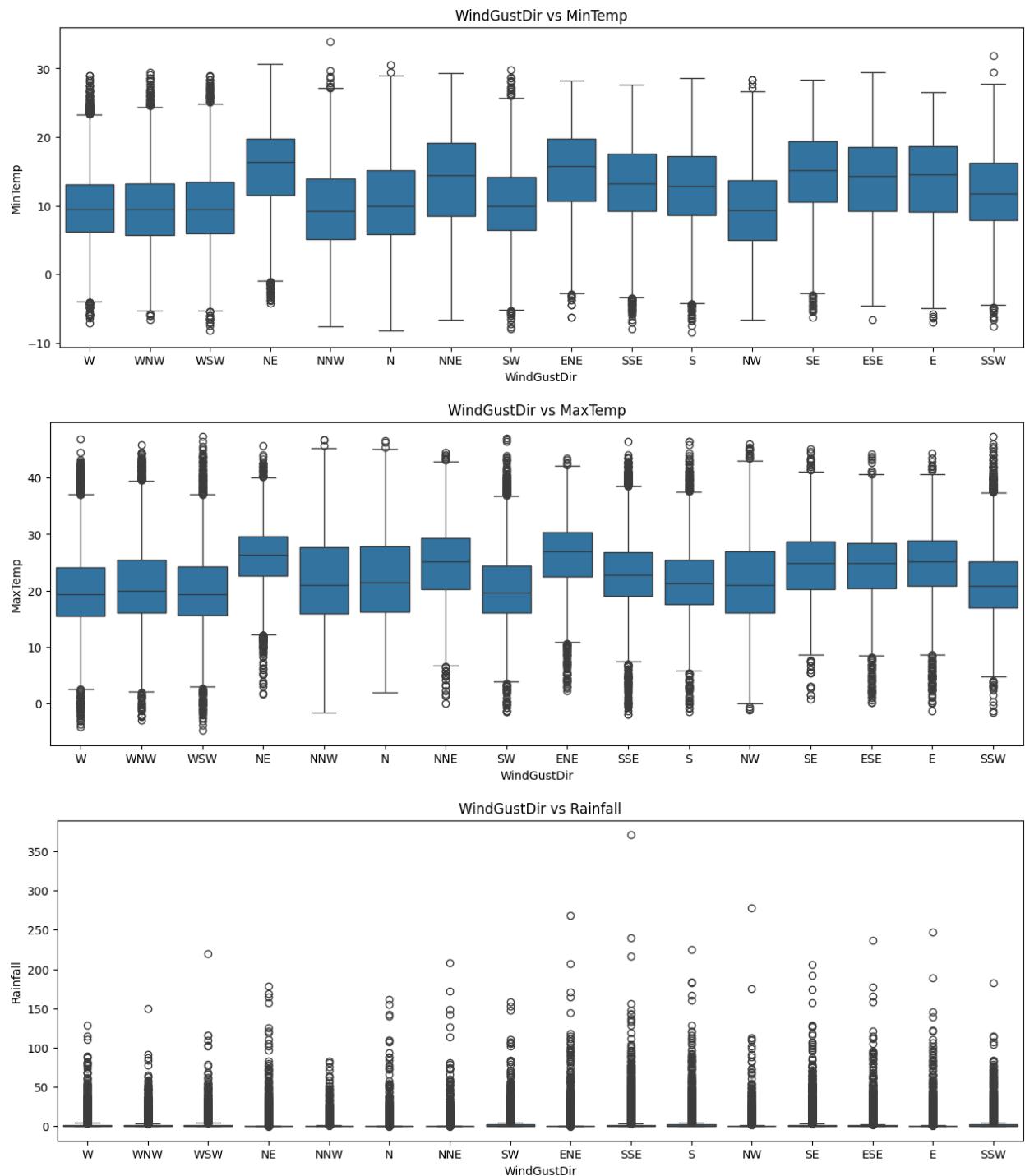
Temp3pm

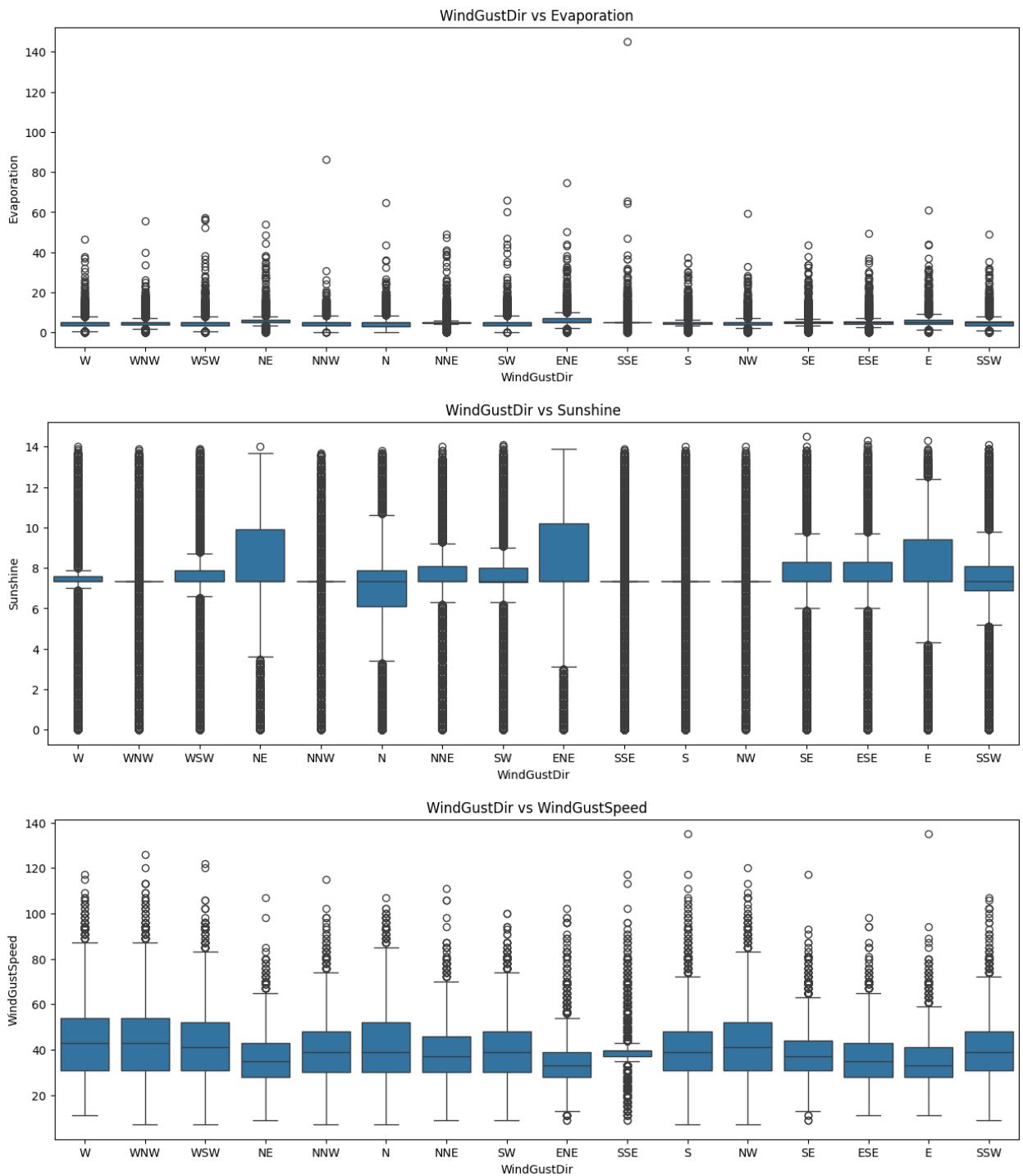


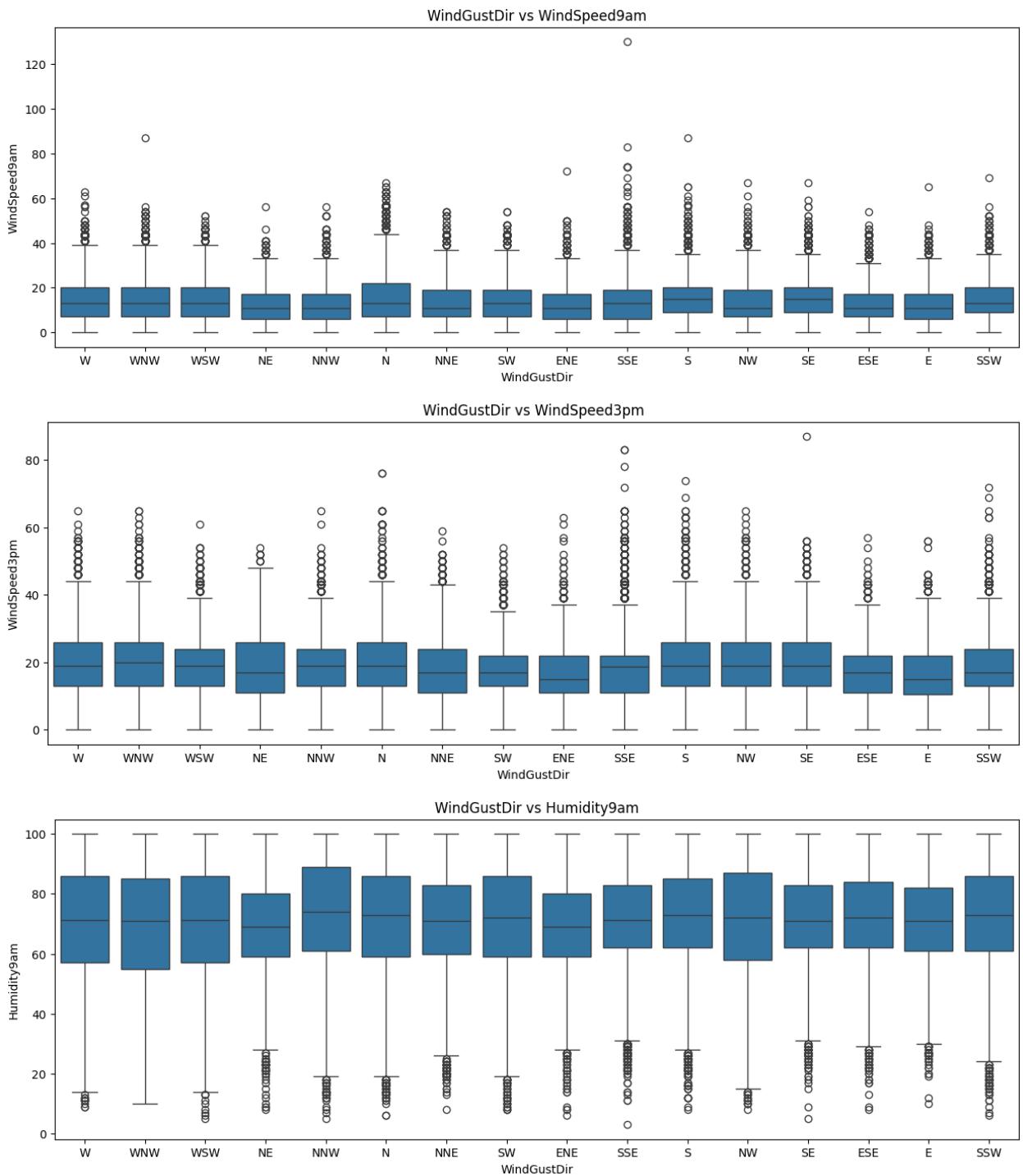
RISK\_MM

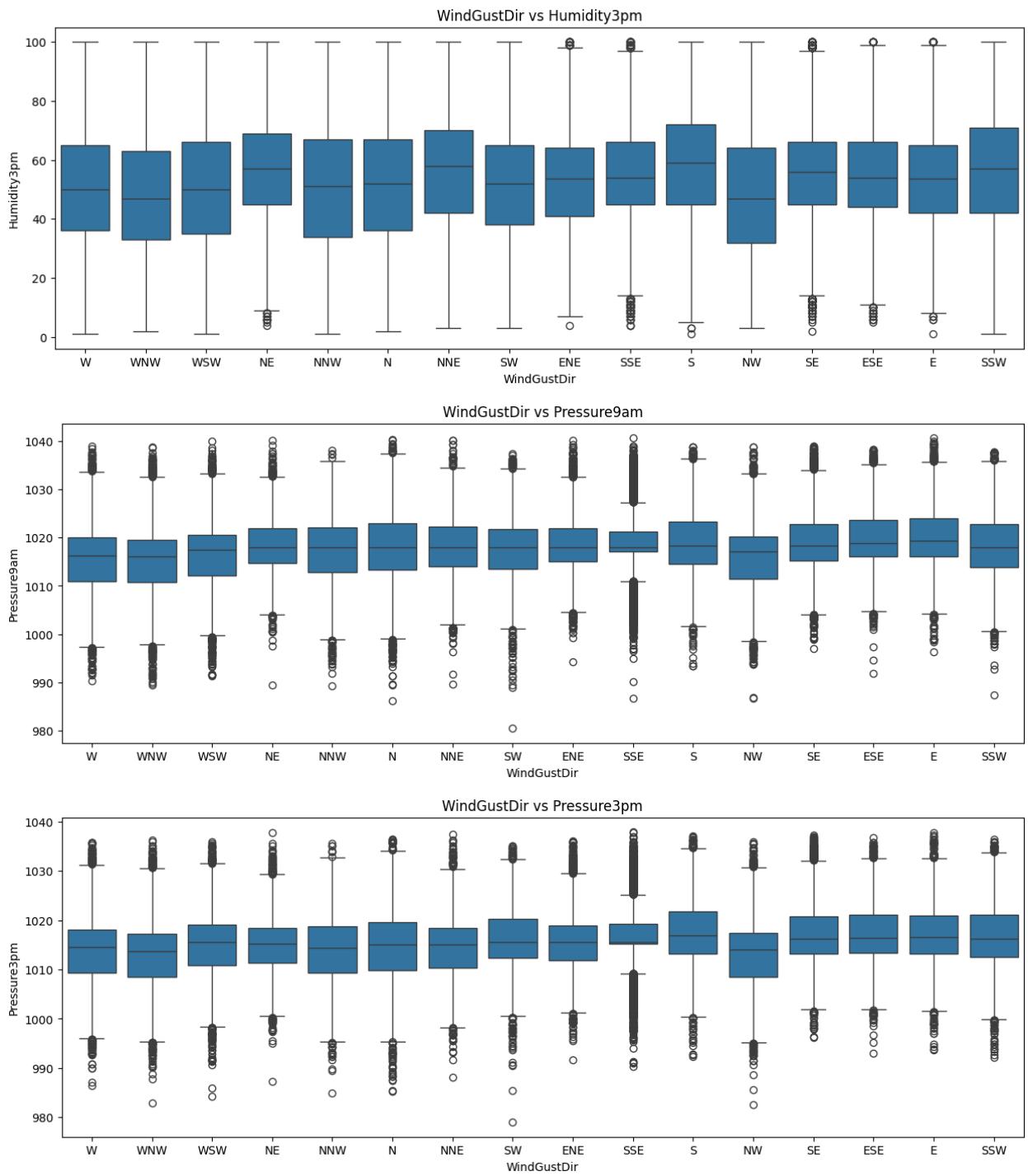


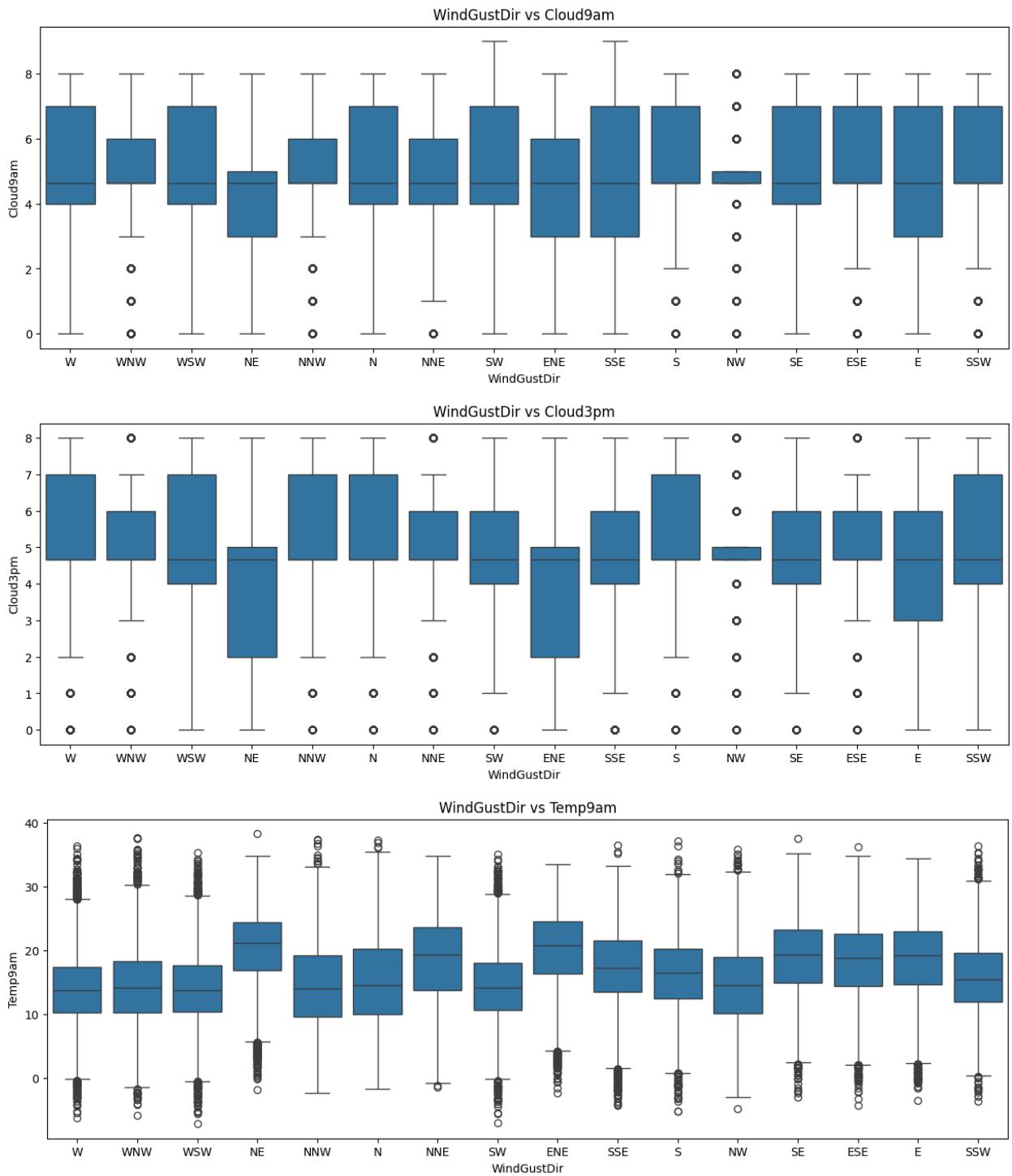
```
In [ ]: for i in categorical5:
    if i!='Date' and i!='Location':
        for j in Numerical5:
            plt.figure(figsize=(15,5))
            sns.boxplot(x=df5[i],y=df5[j])
            plt.title(i+' vs '+j)
            plt.show()
```

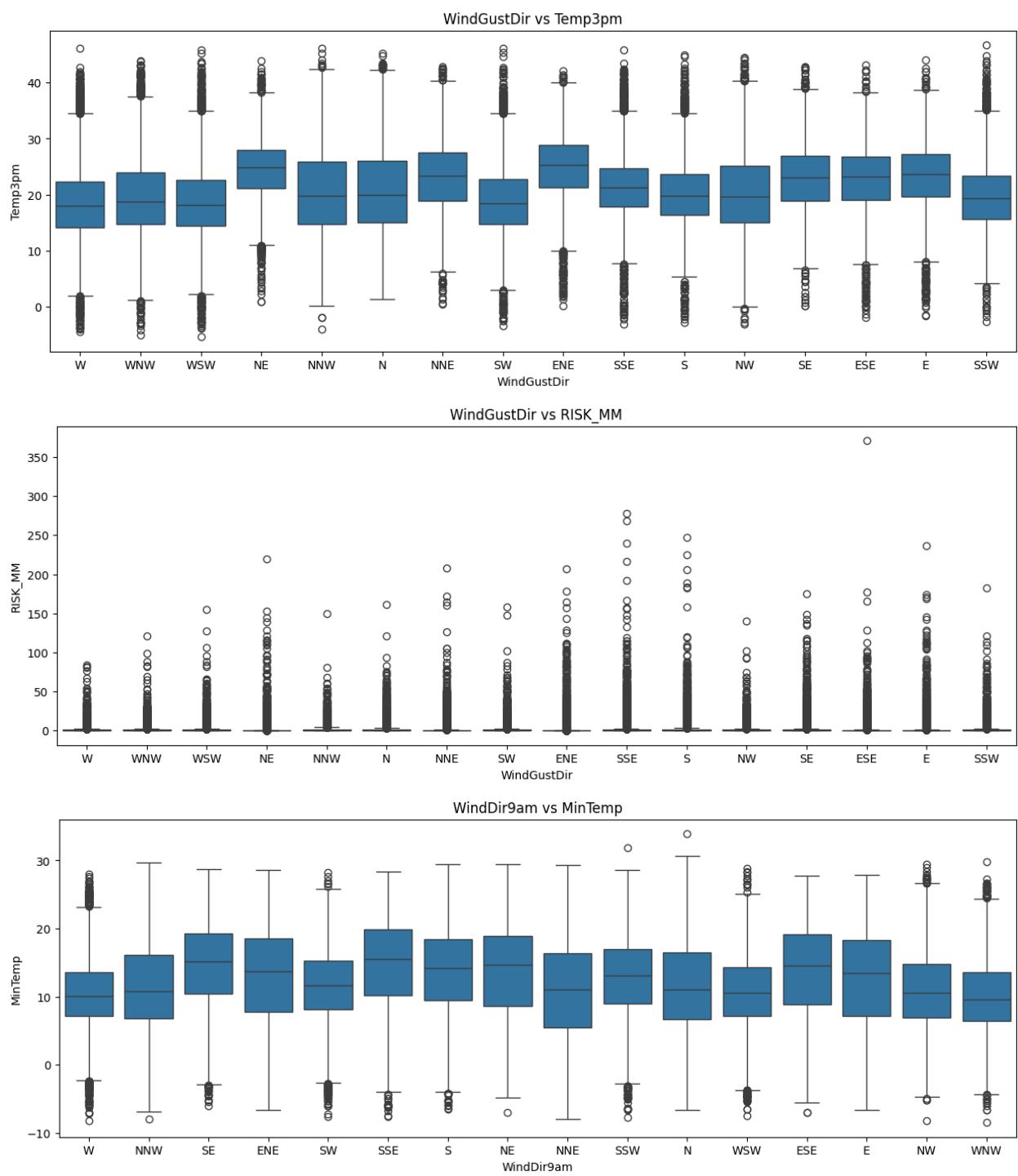




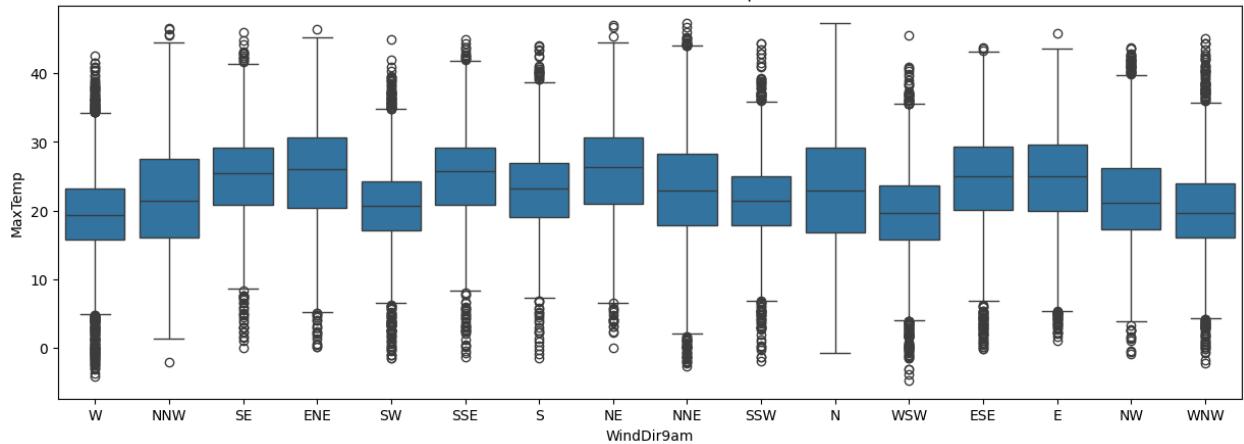




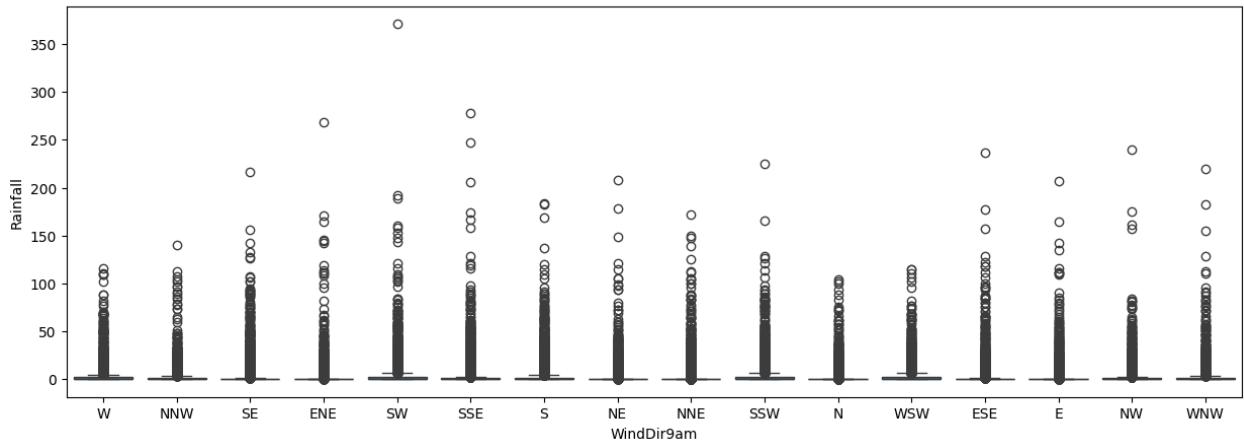




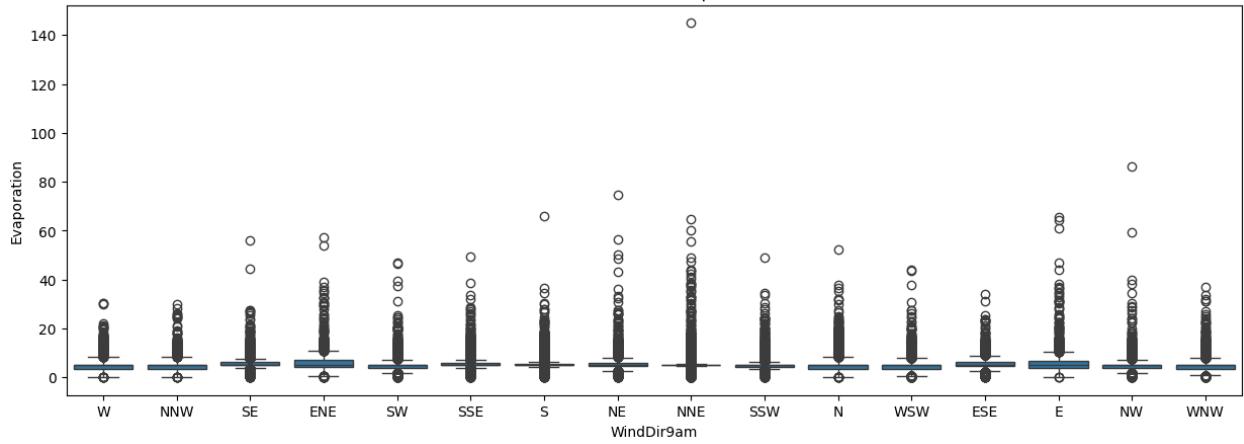
WindDir9am vs MaxTemp

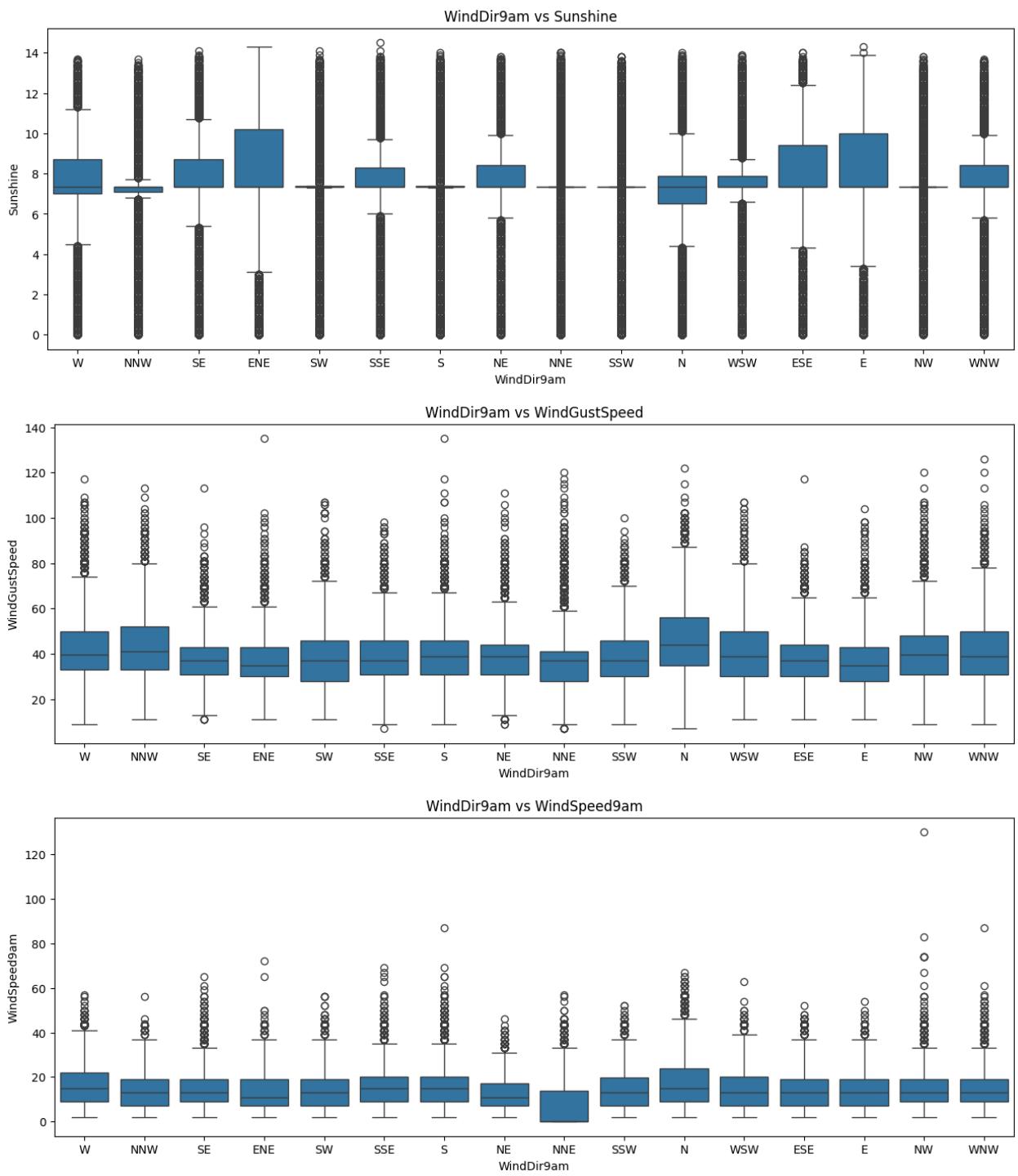


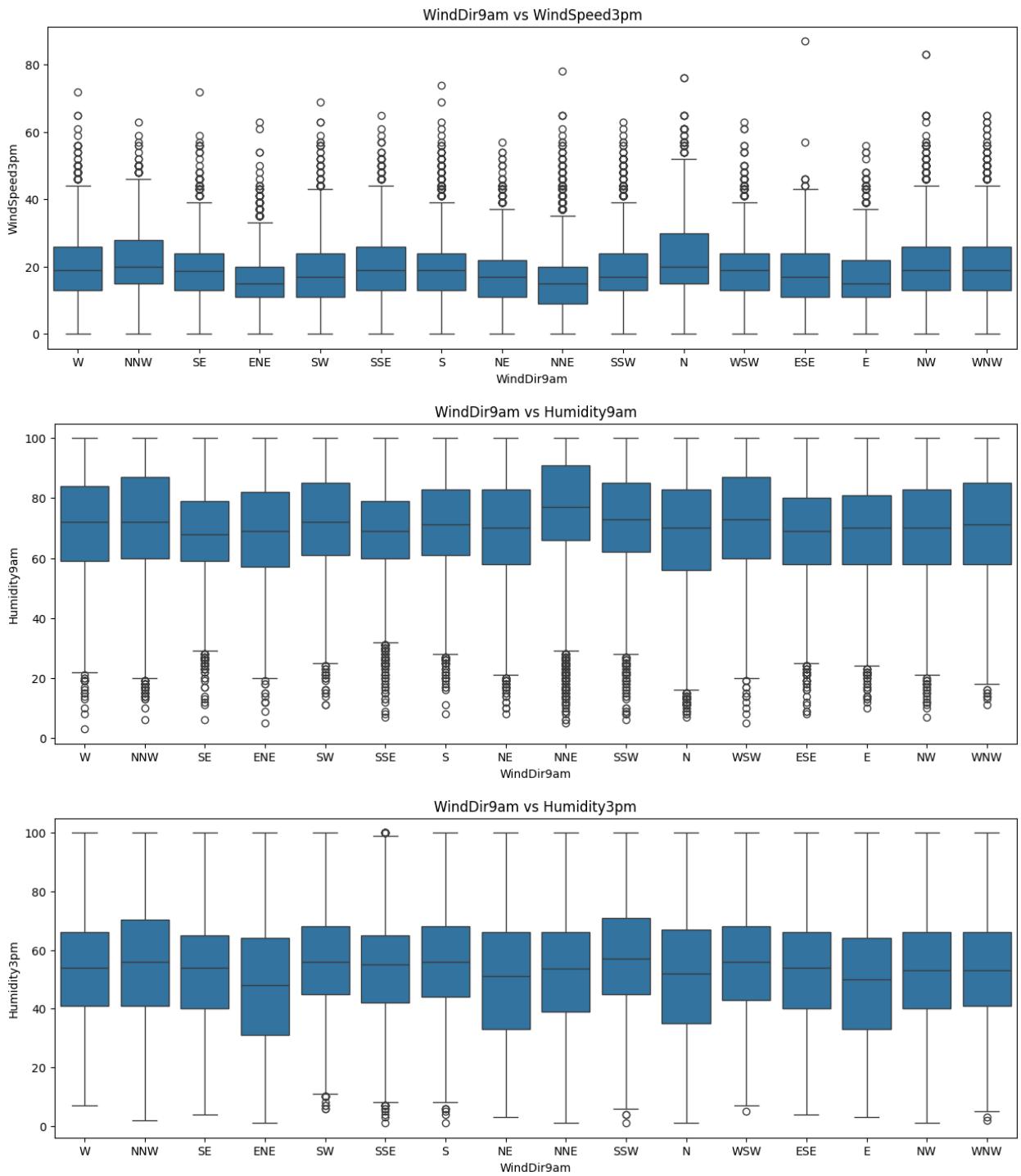
WindDir9am vs Rainfall

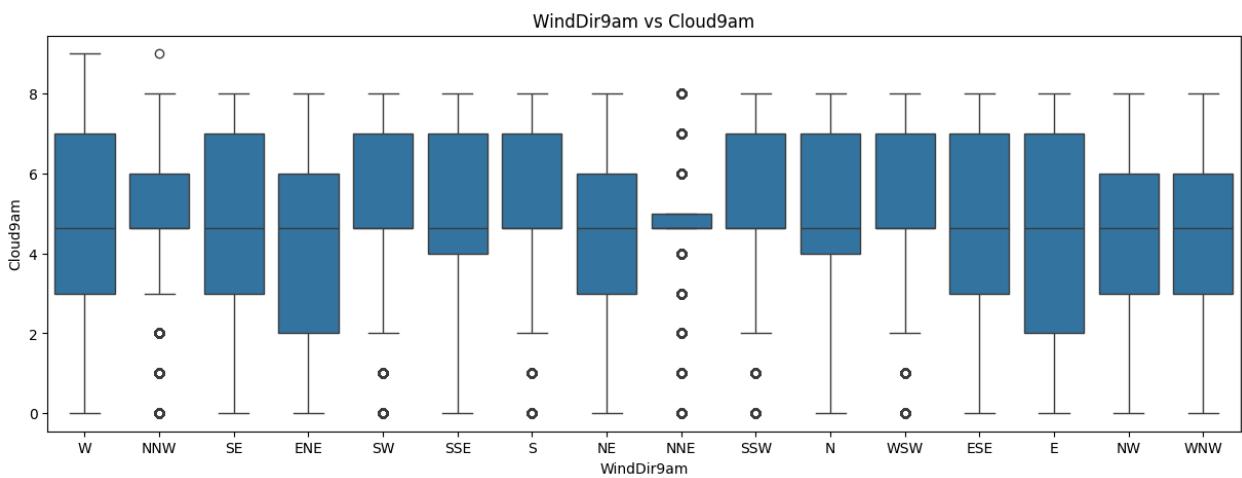
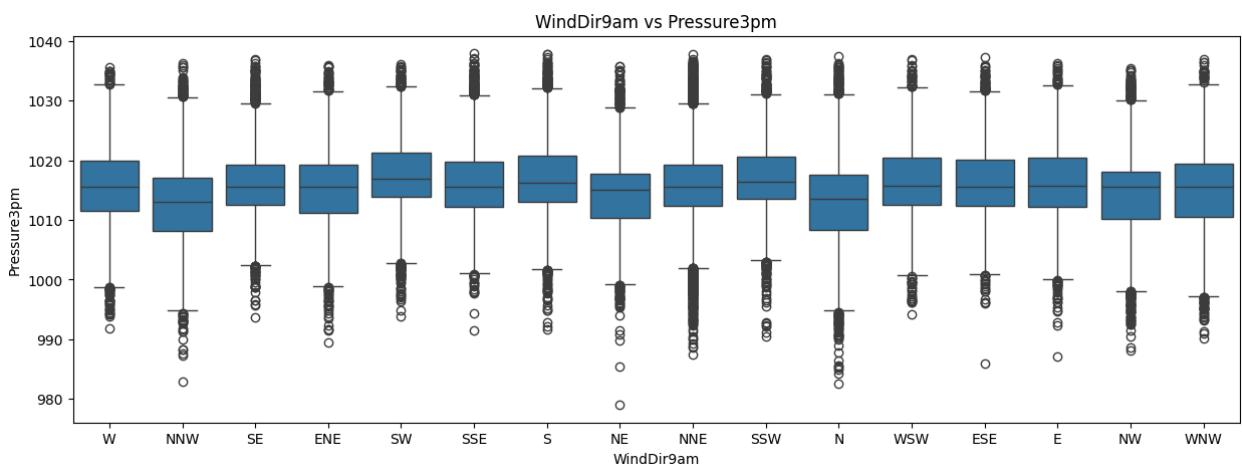
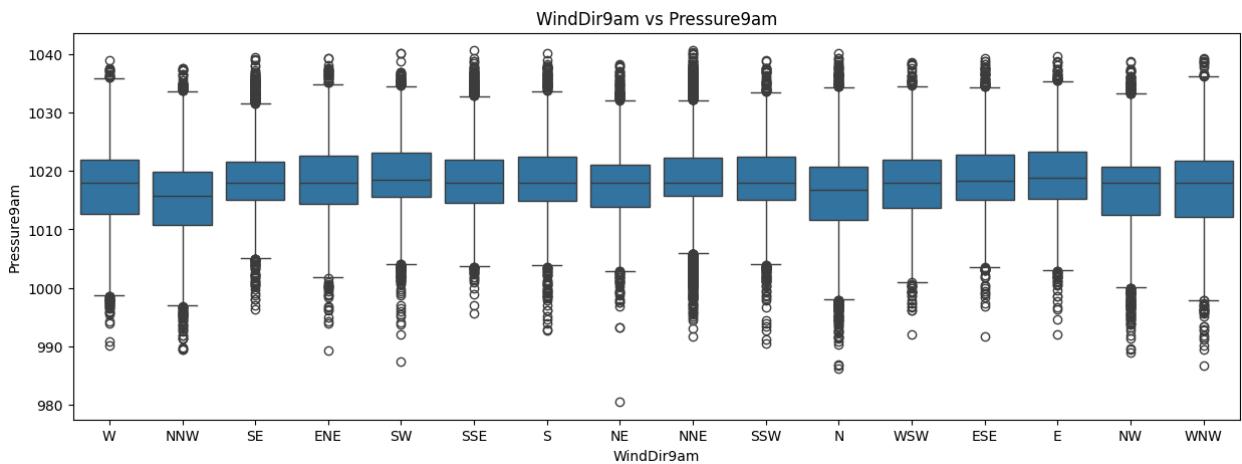


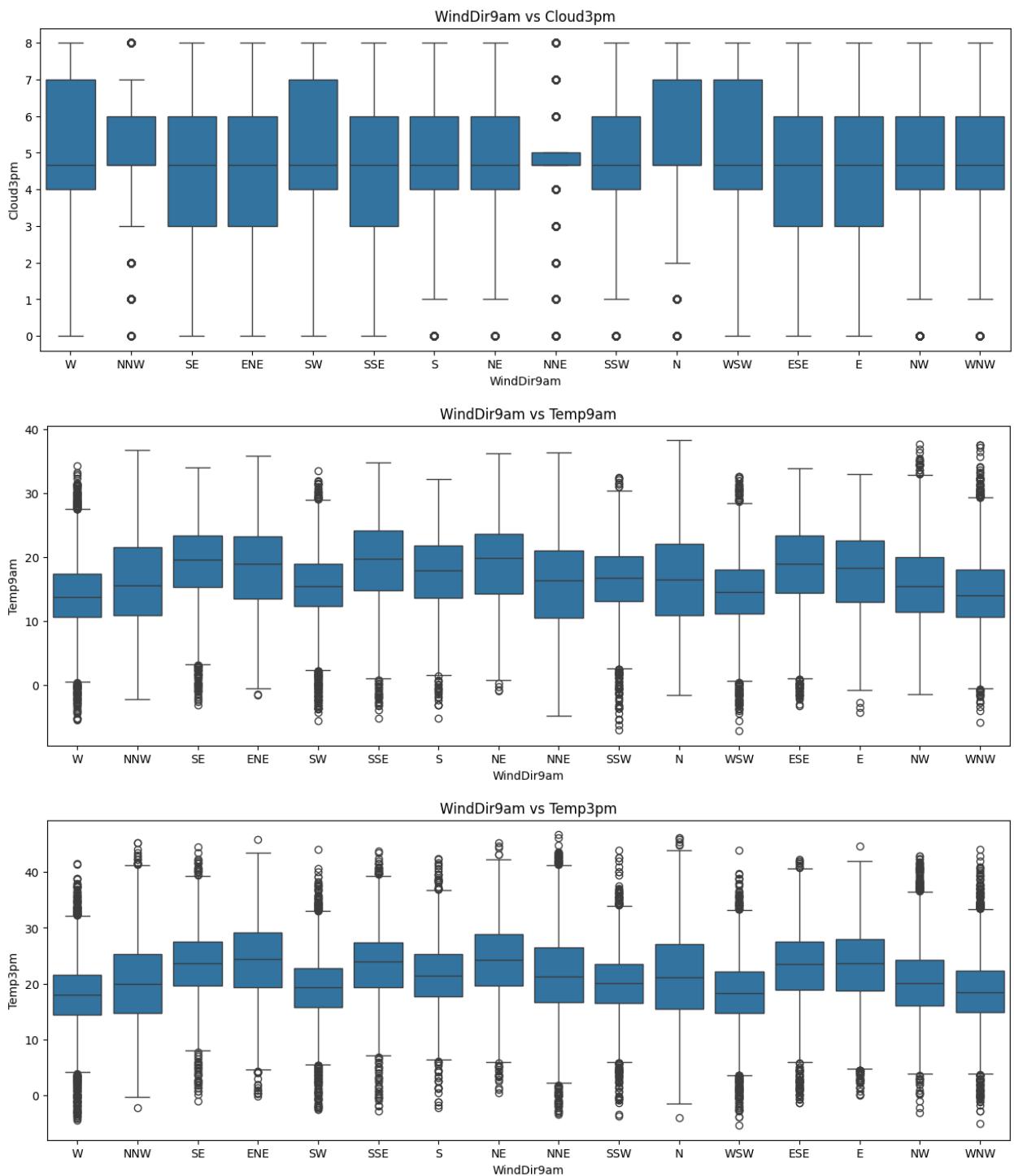
WindDir9am vs Evaporation

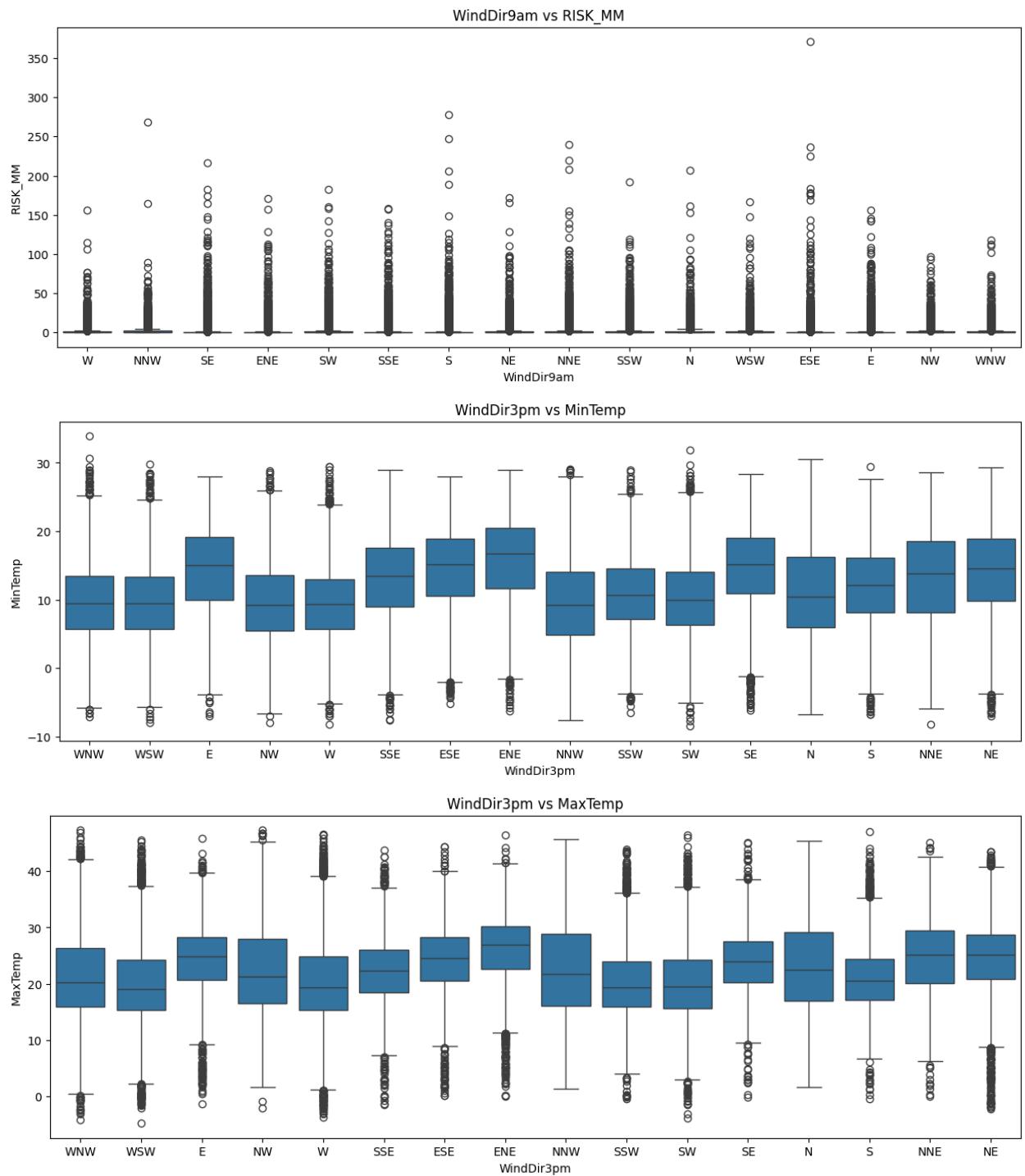


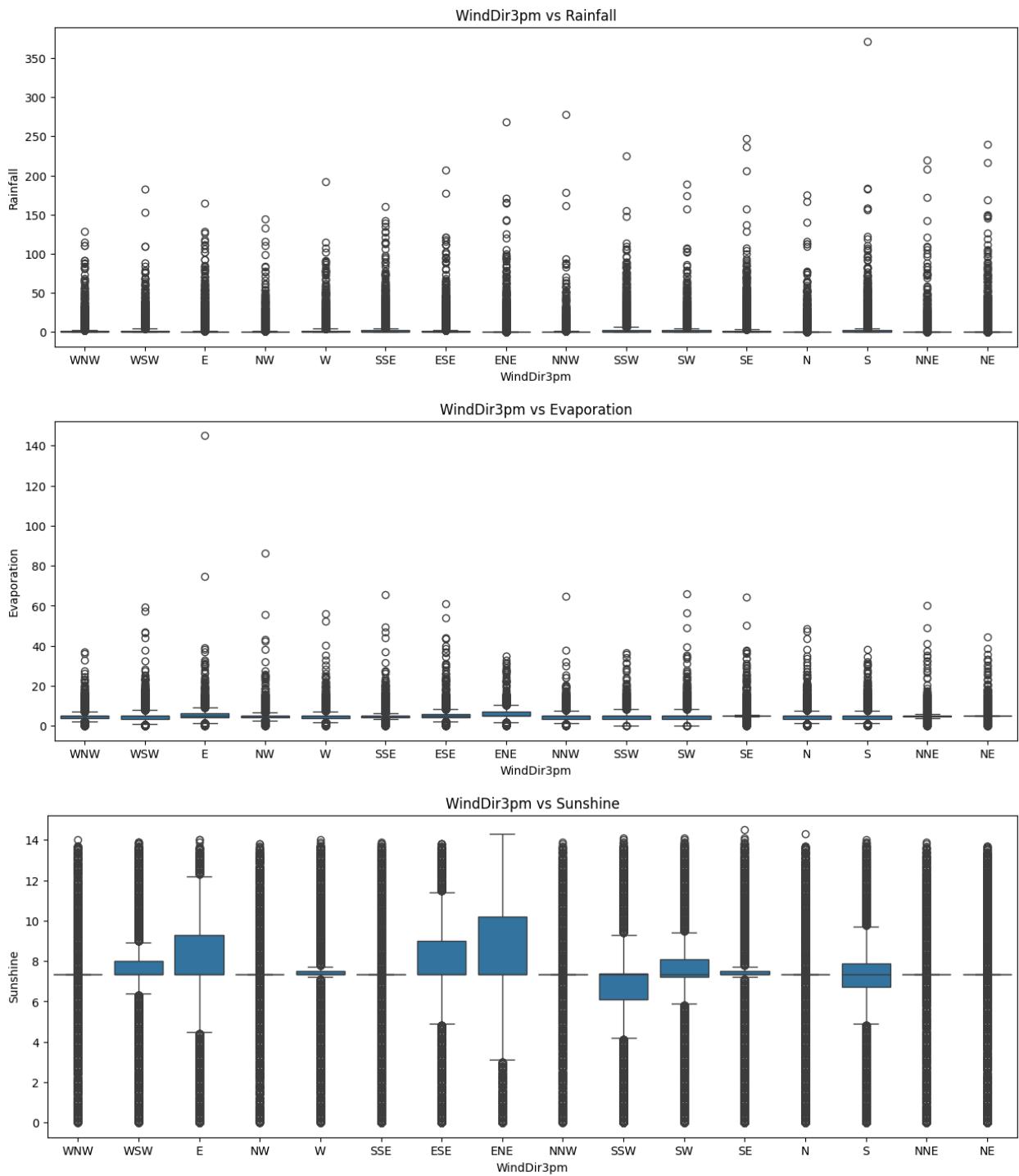


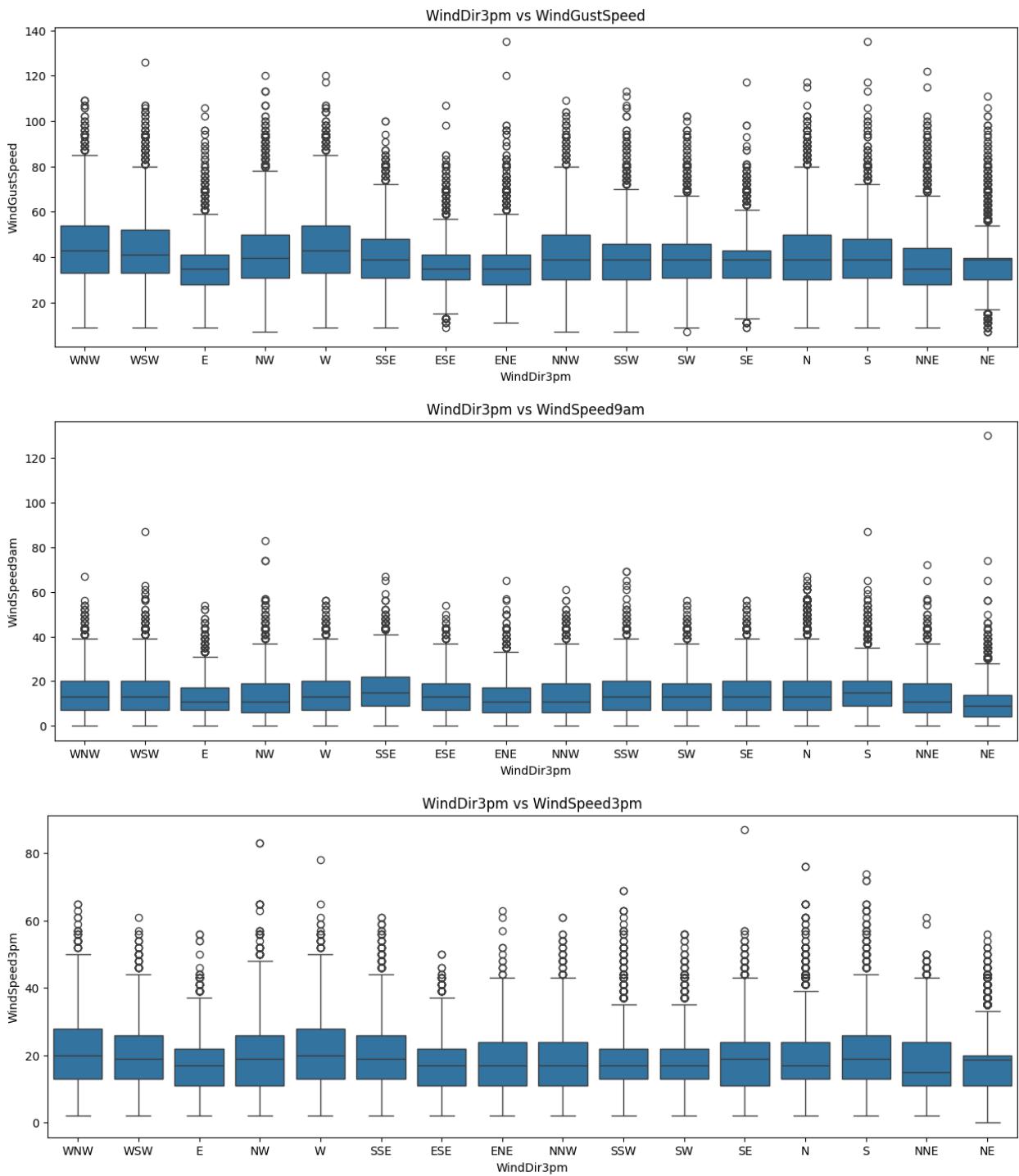


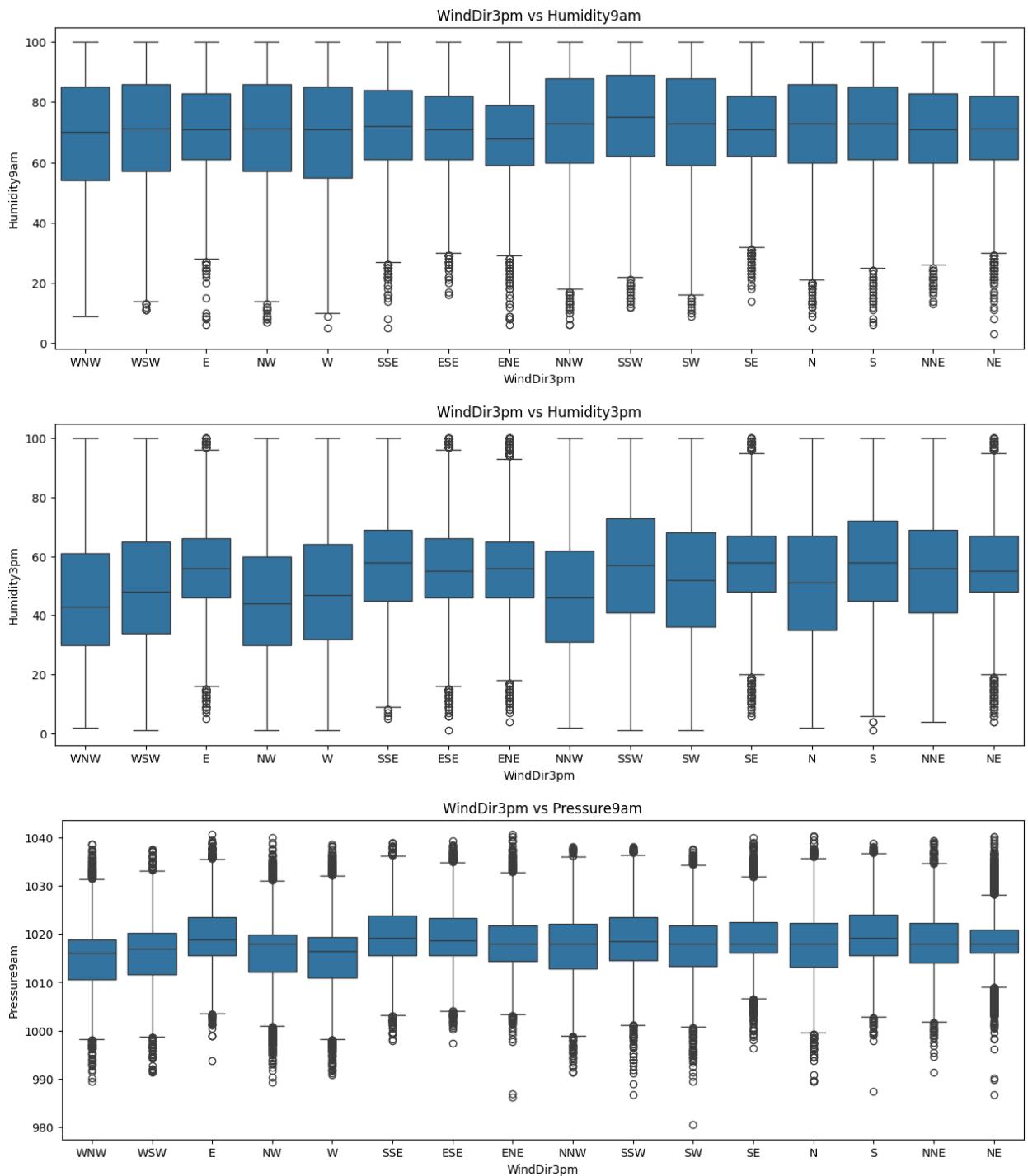


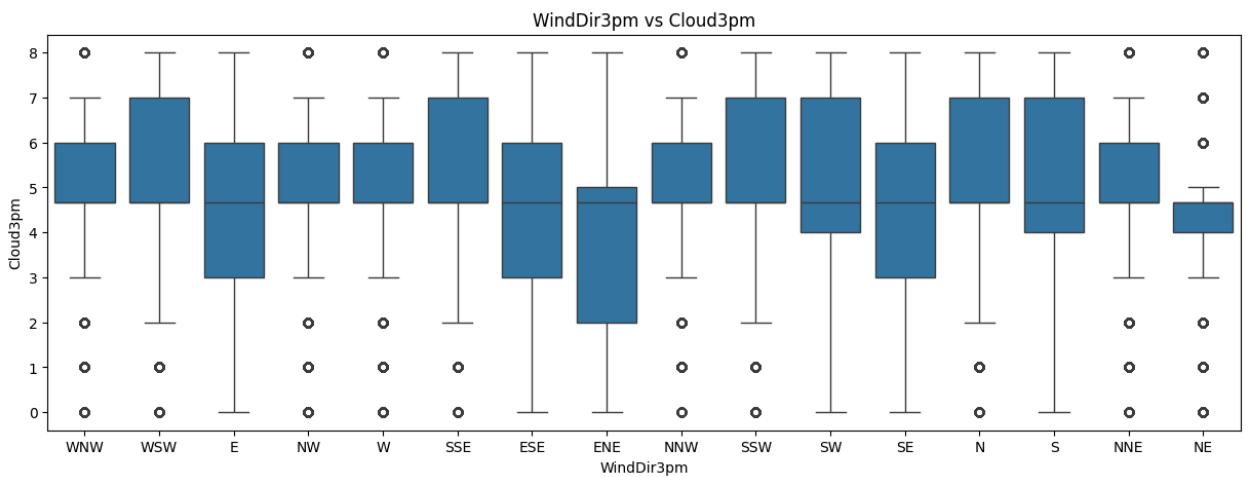
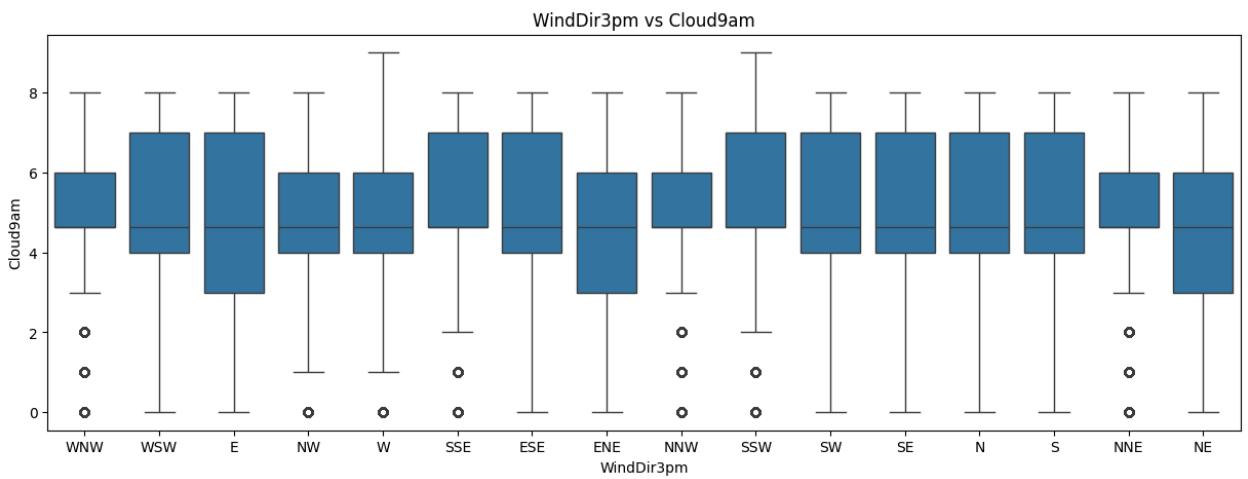
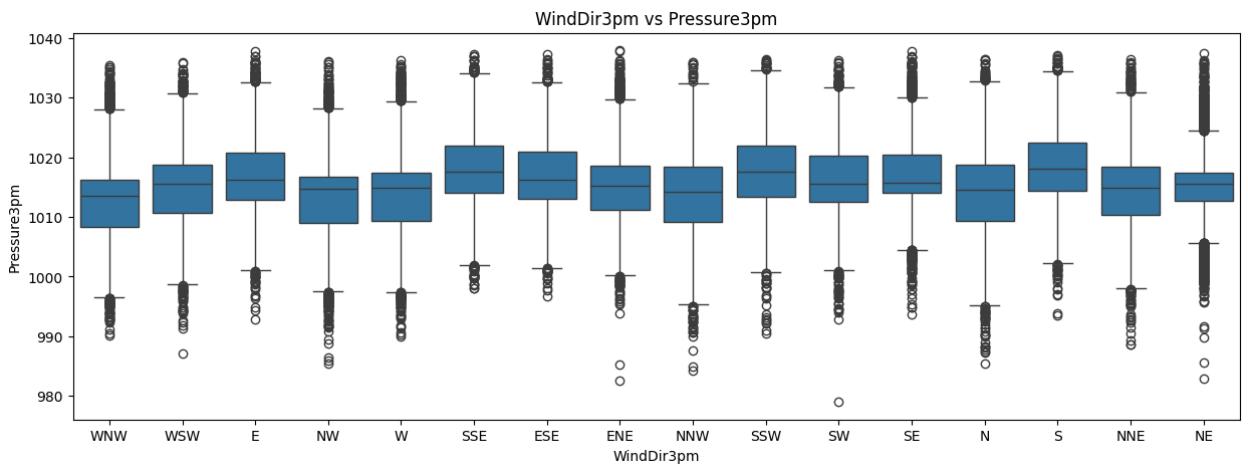


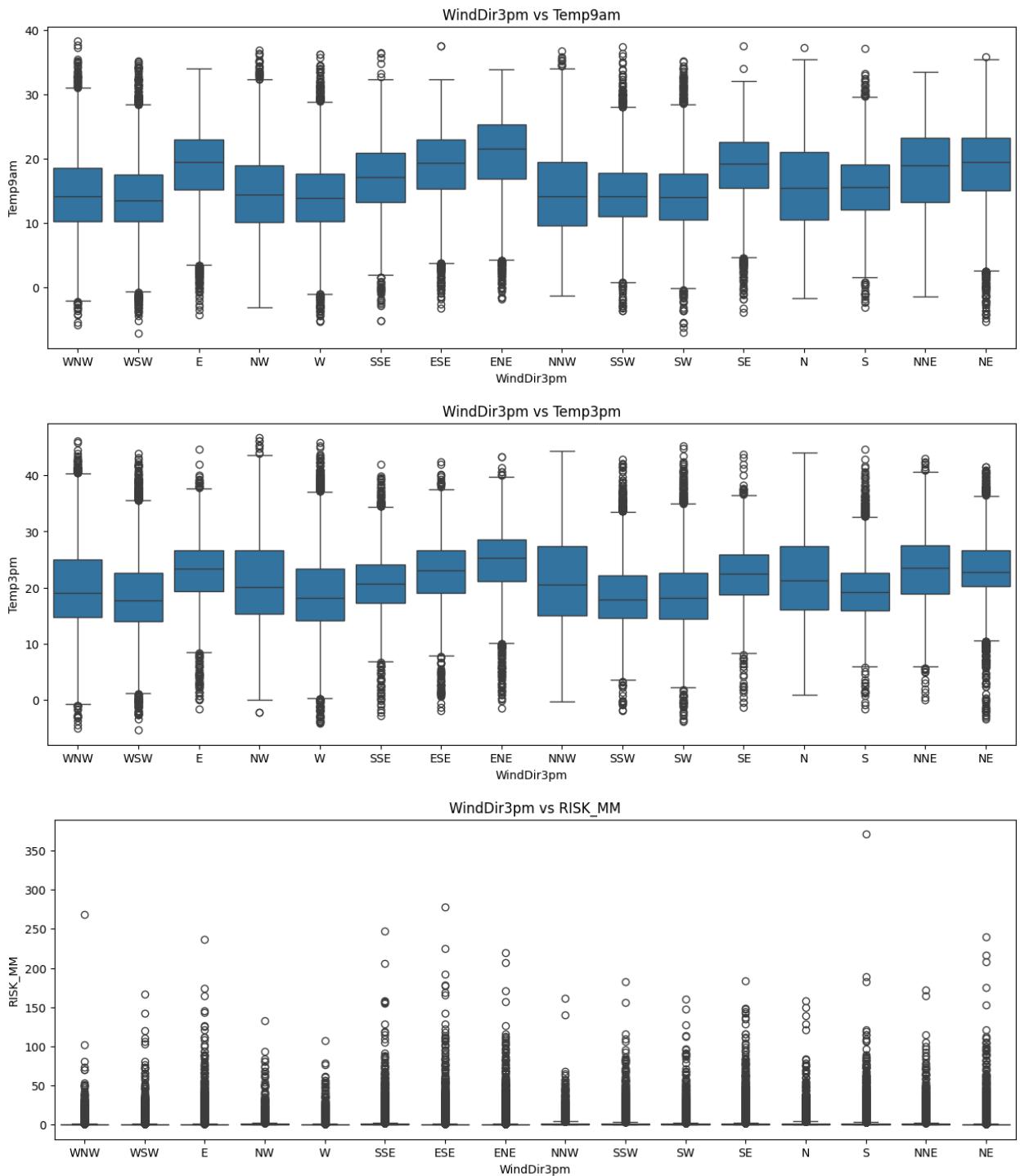


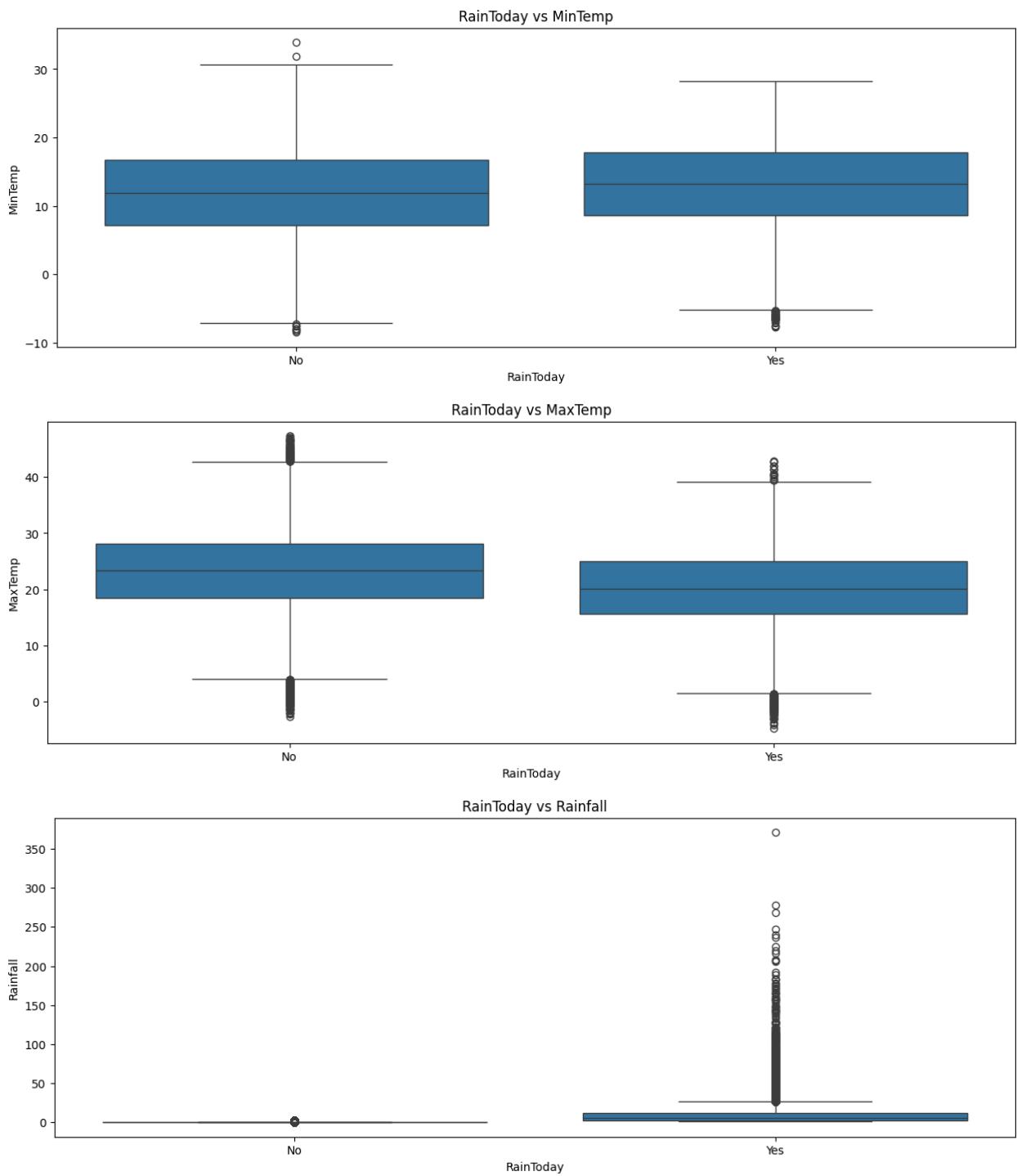


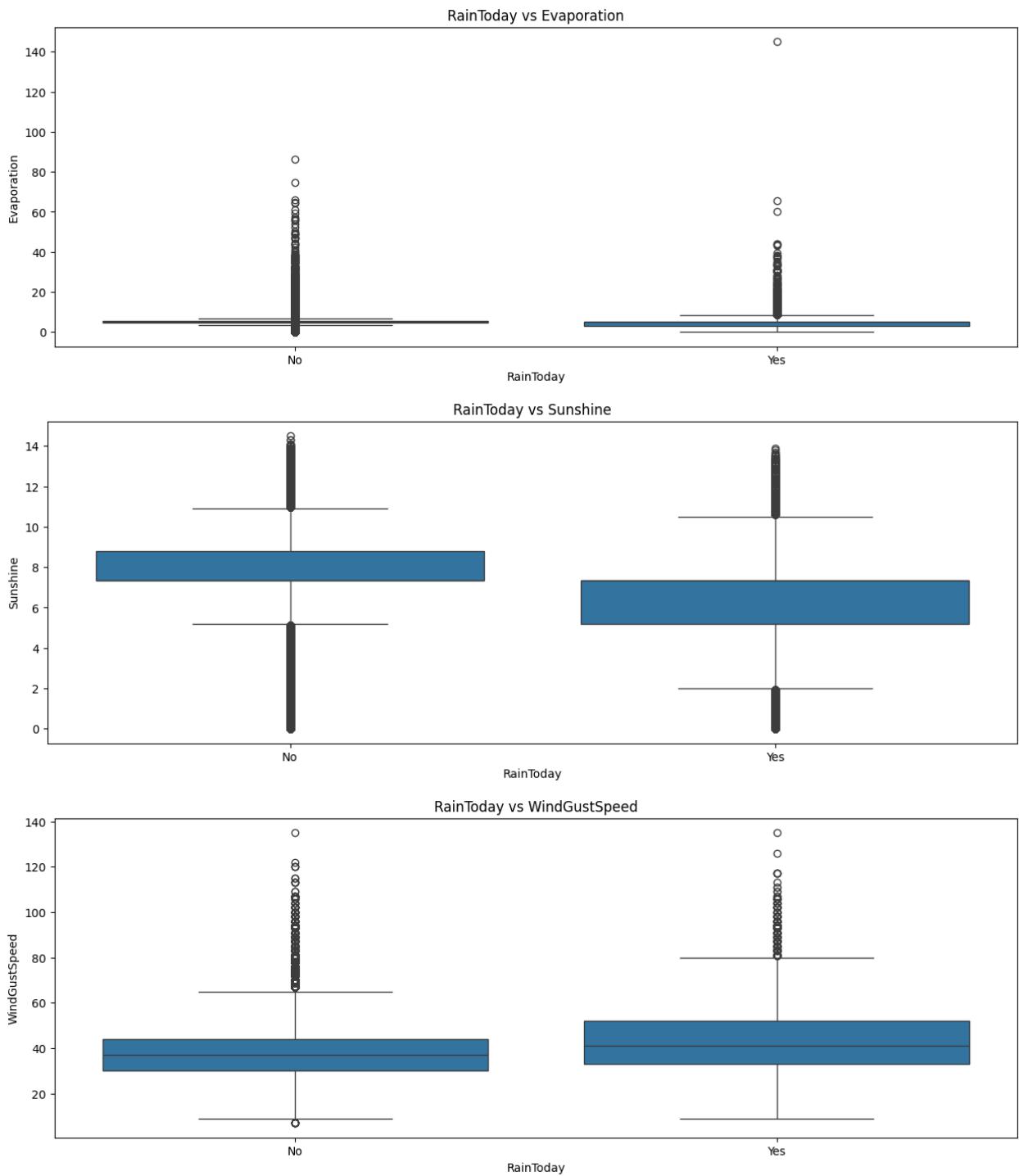


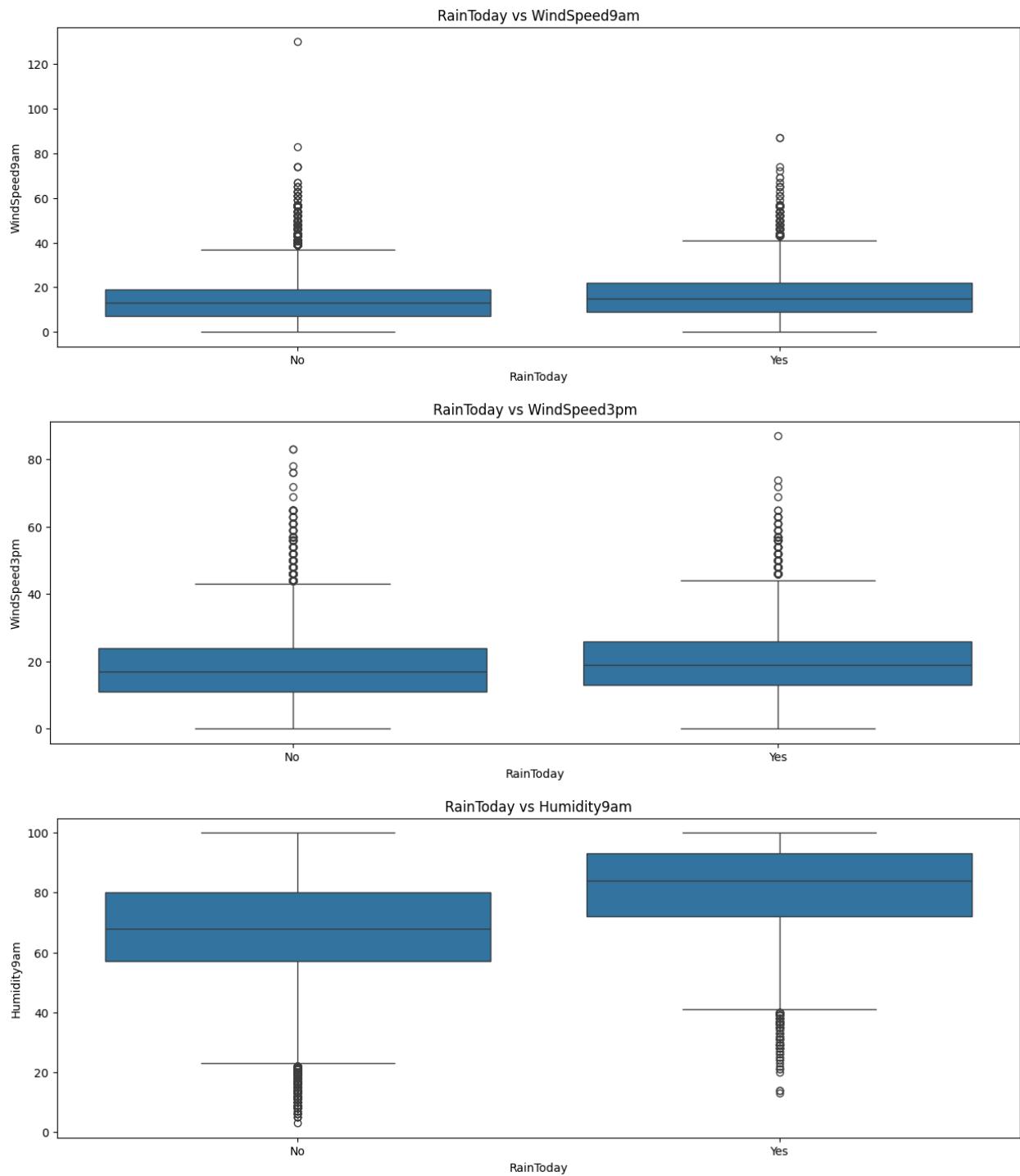


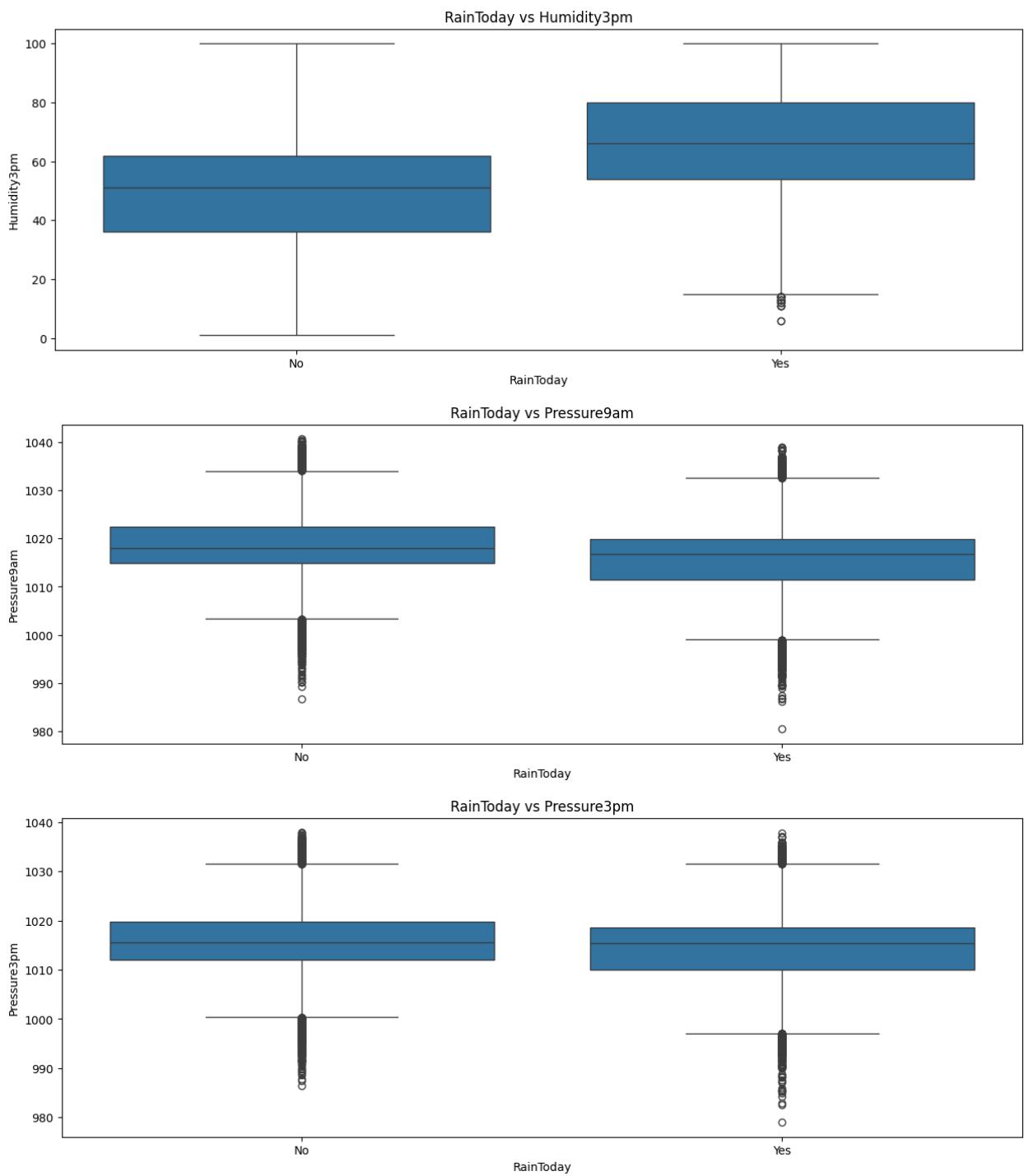


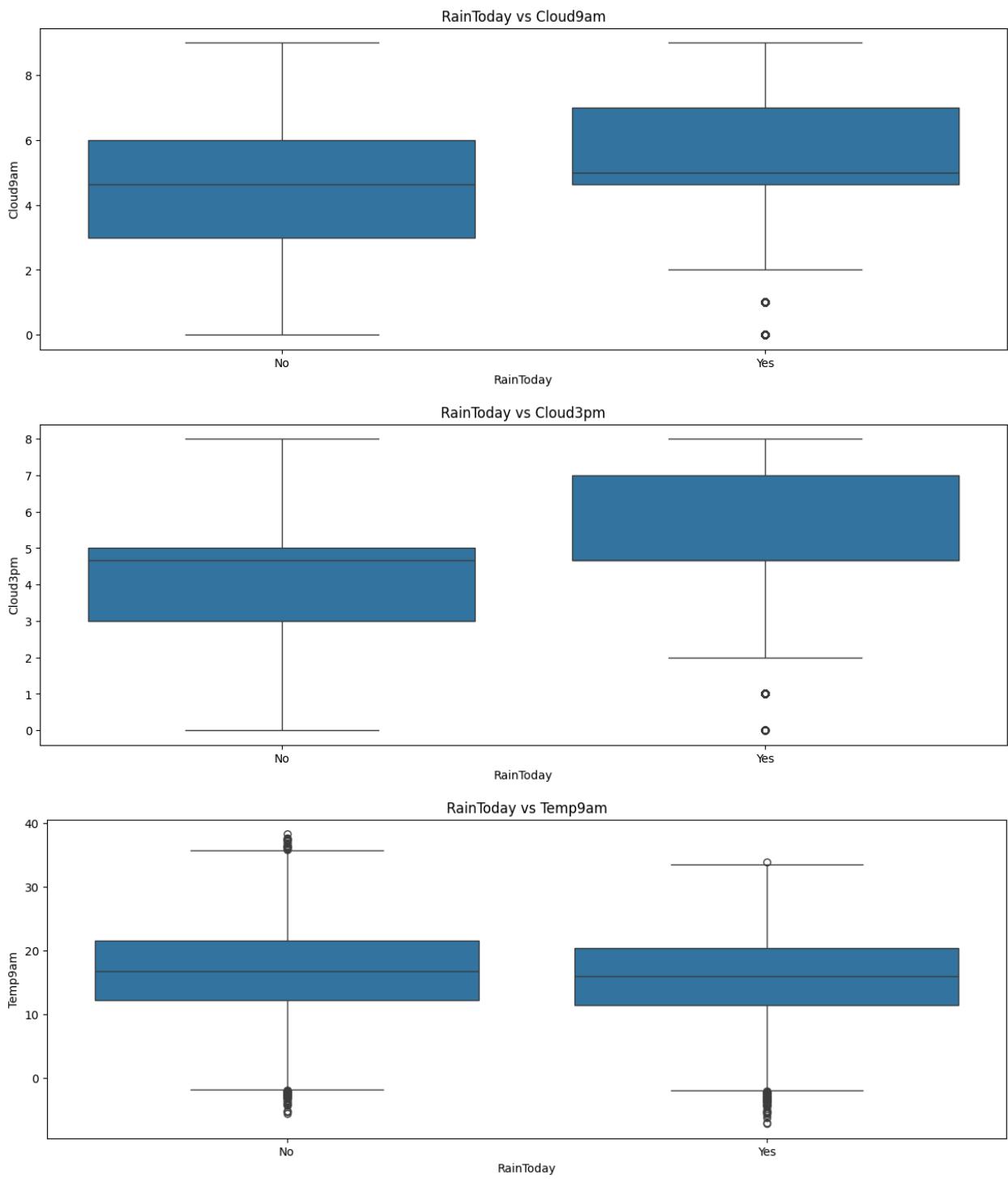


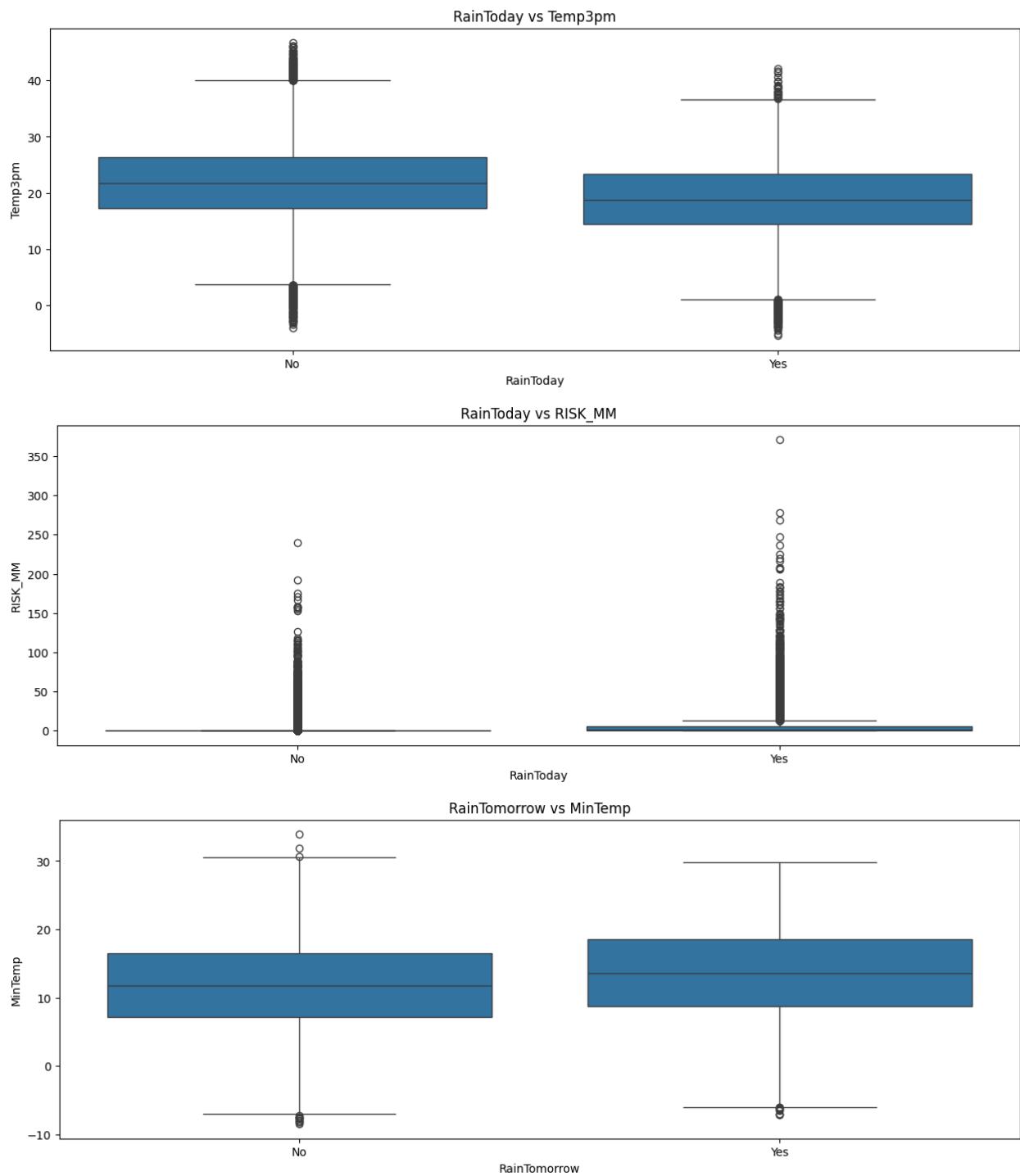




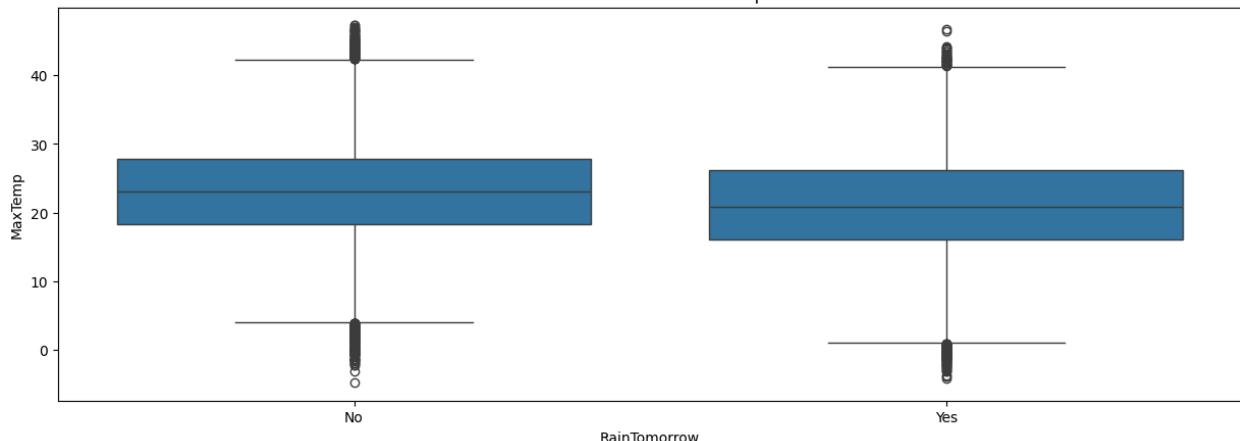




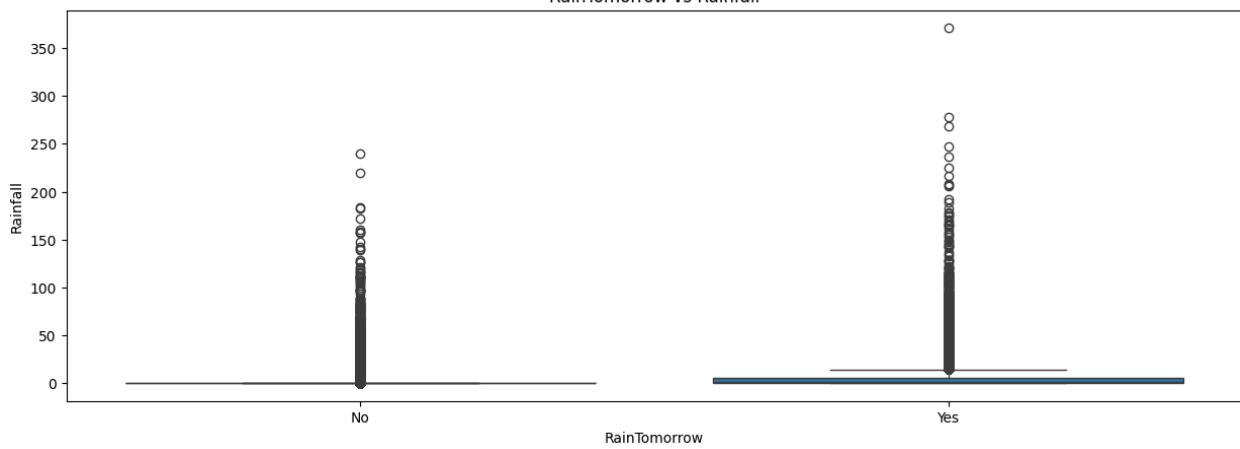




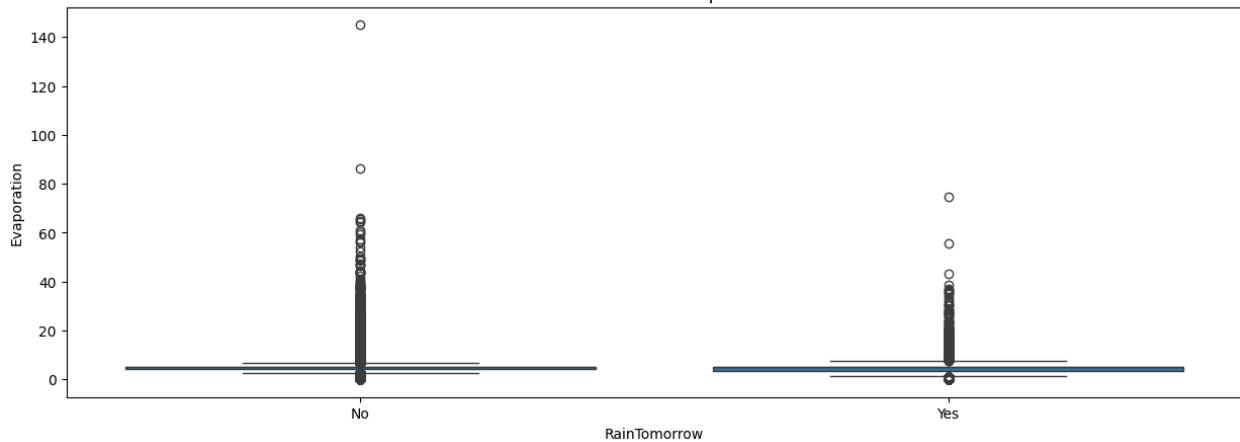
RainTomorrow vs MaxTemp

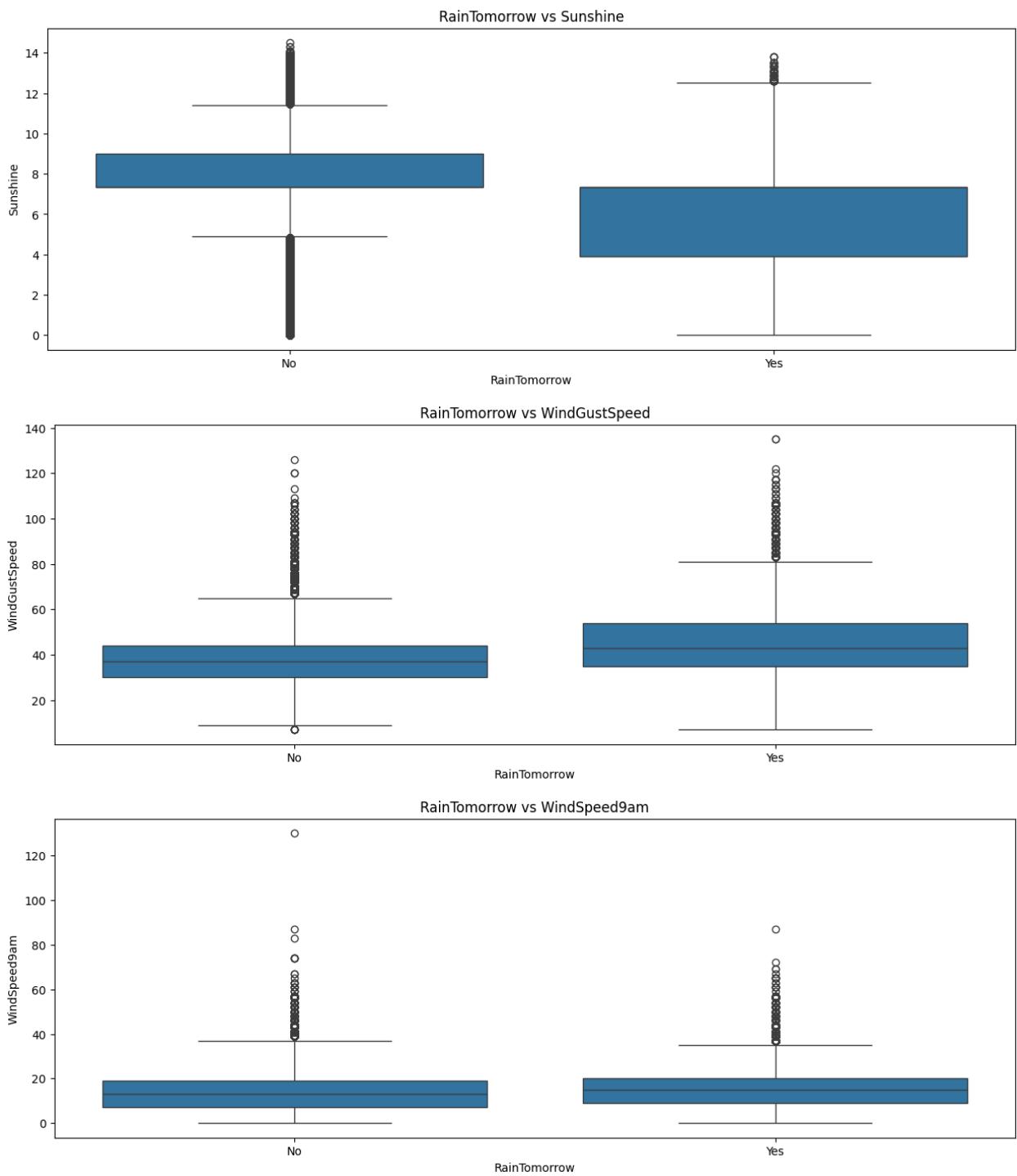


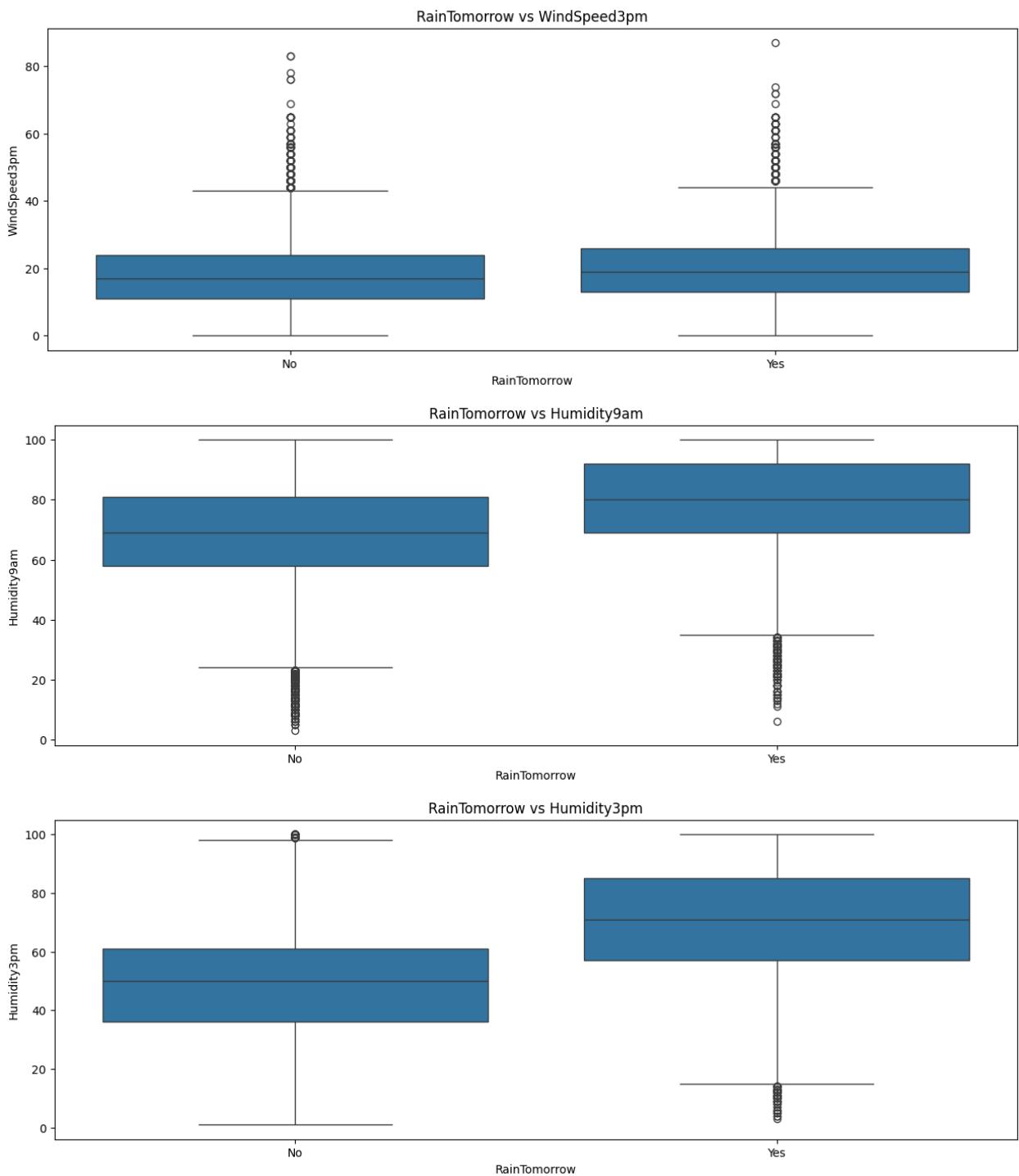
RainTomorrow vs Rainfall

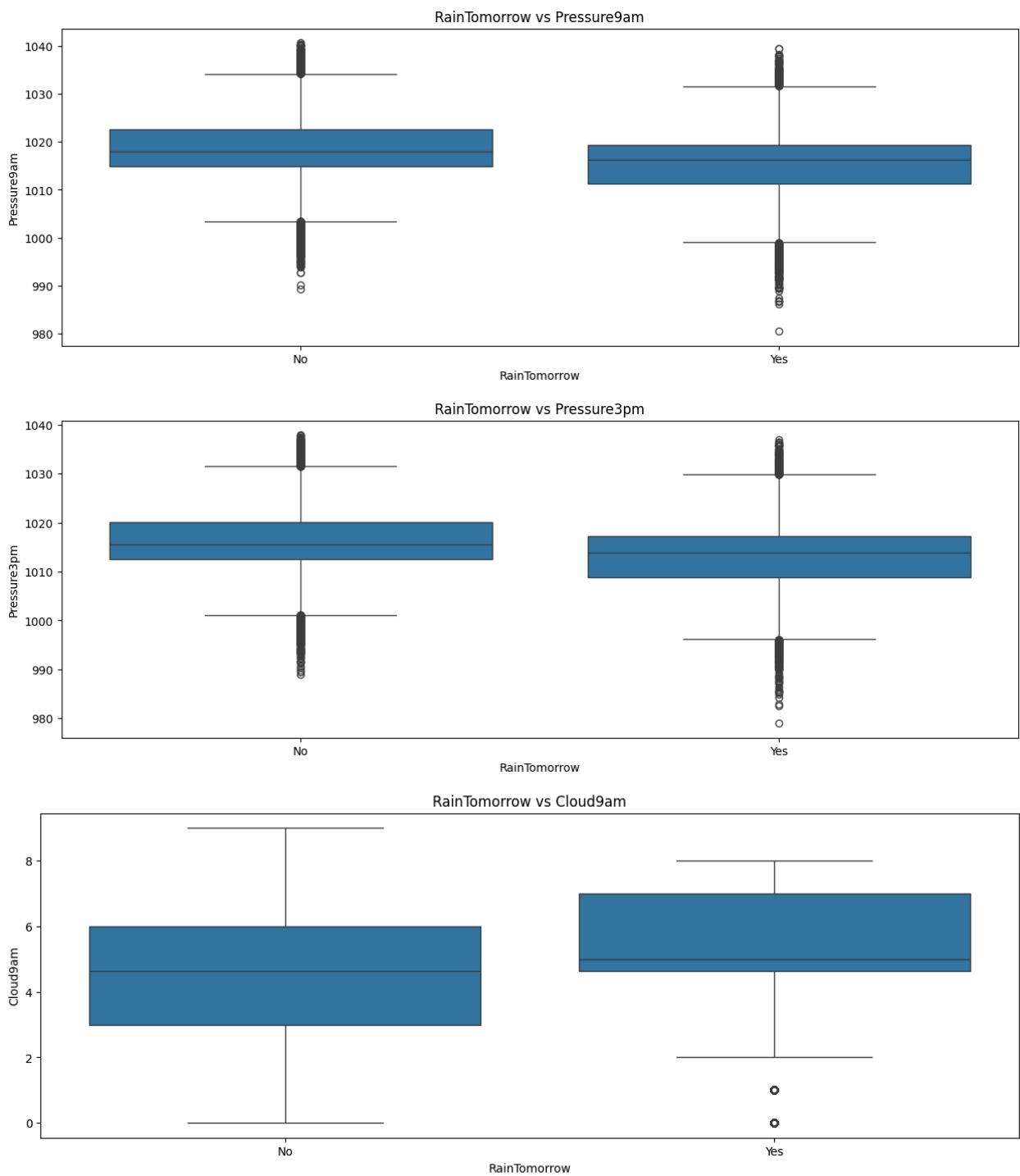


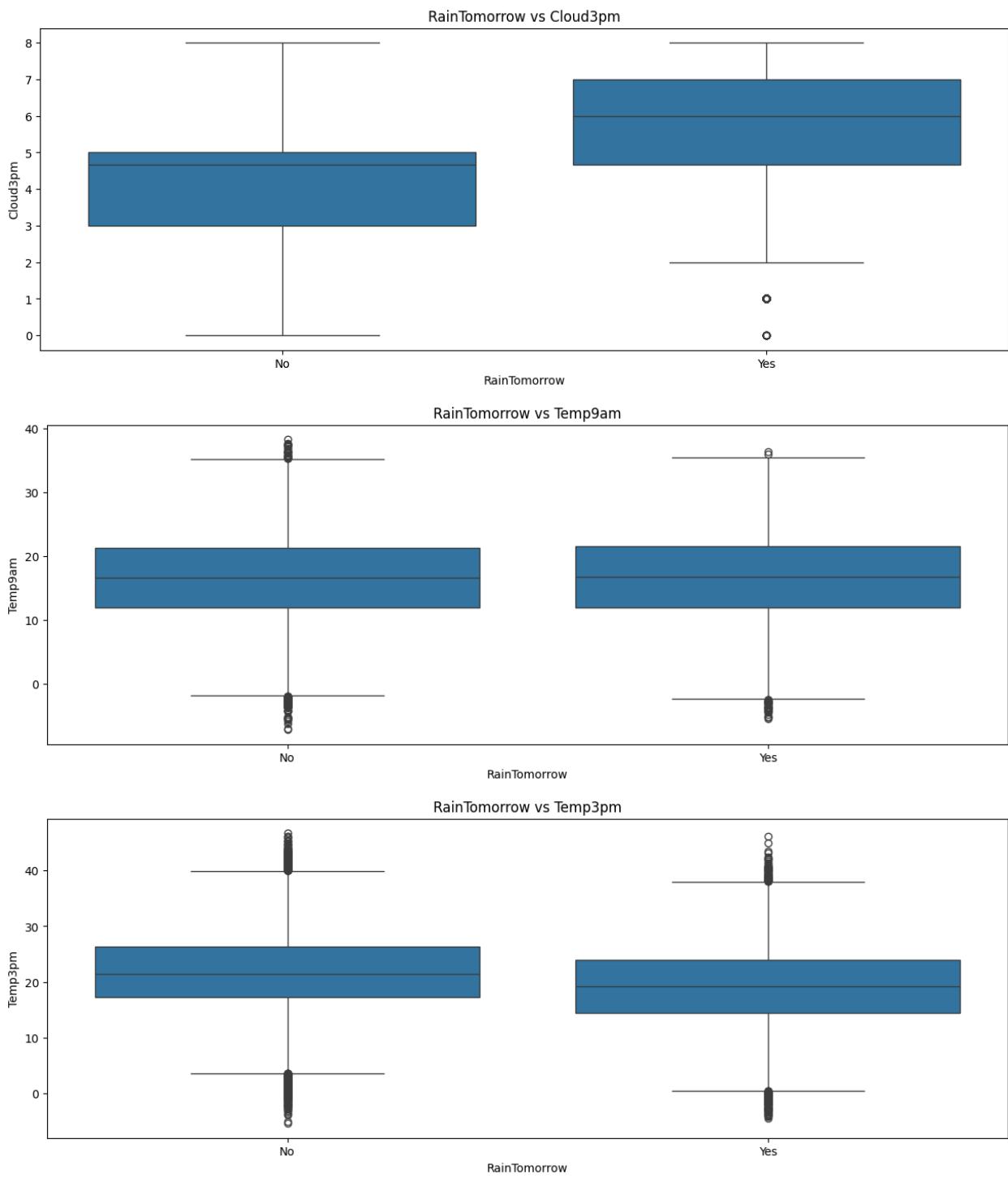
RainTomorrow vs Evaporation

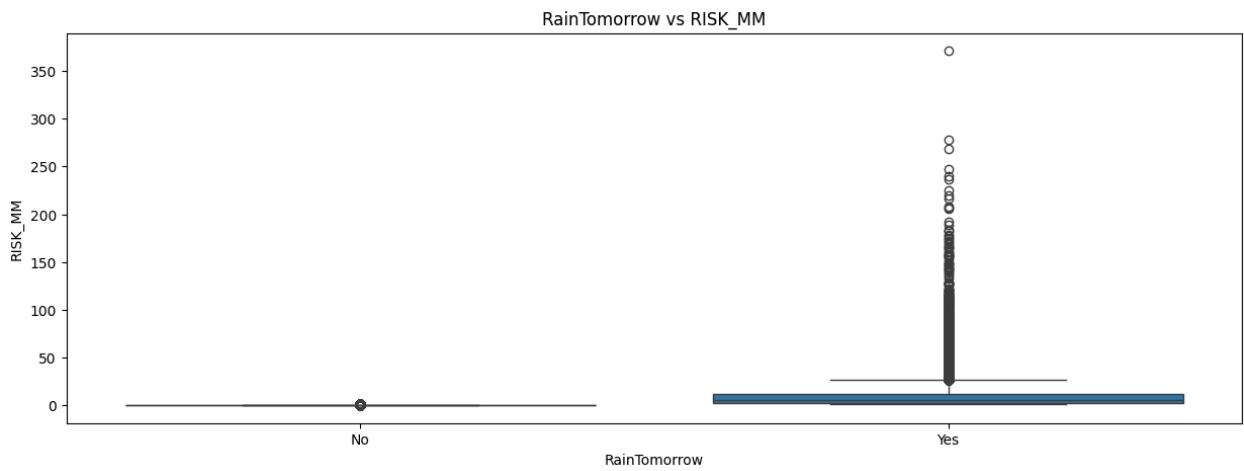




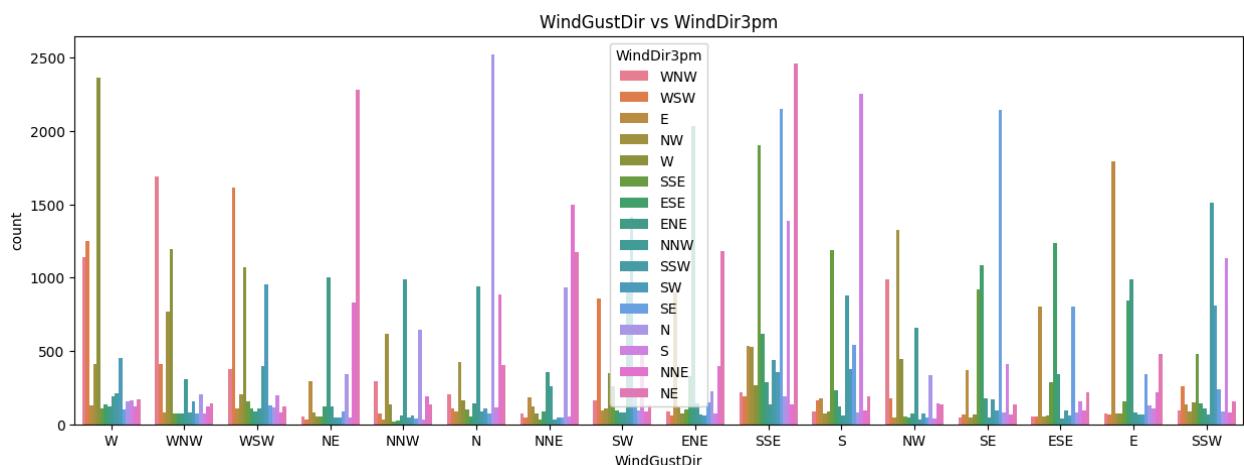
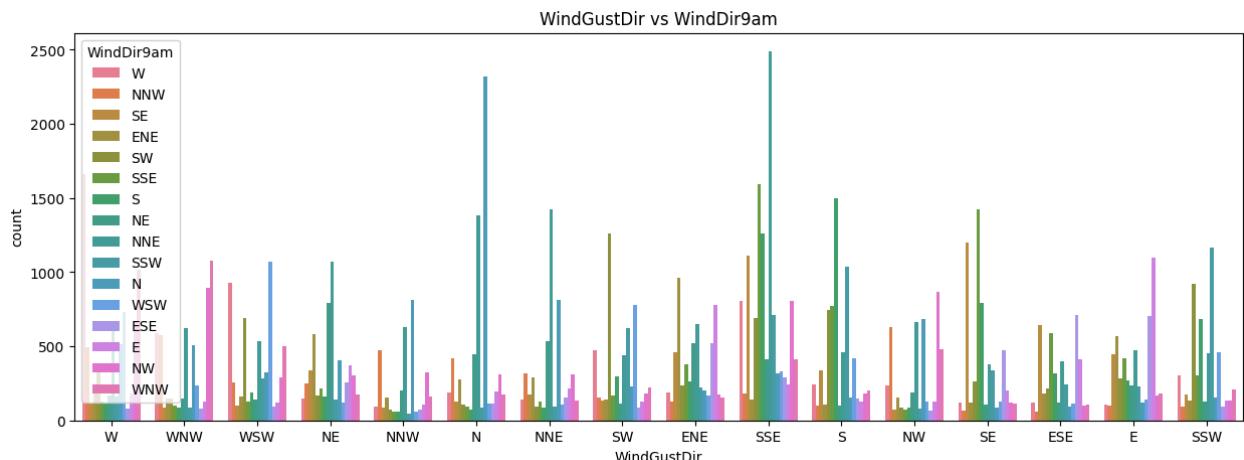


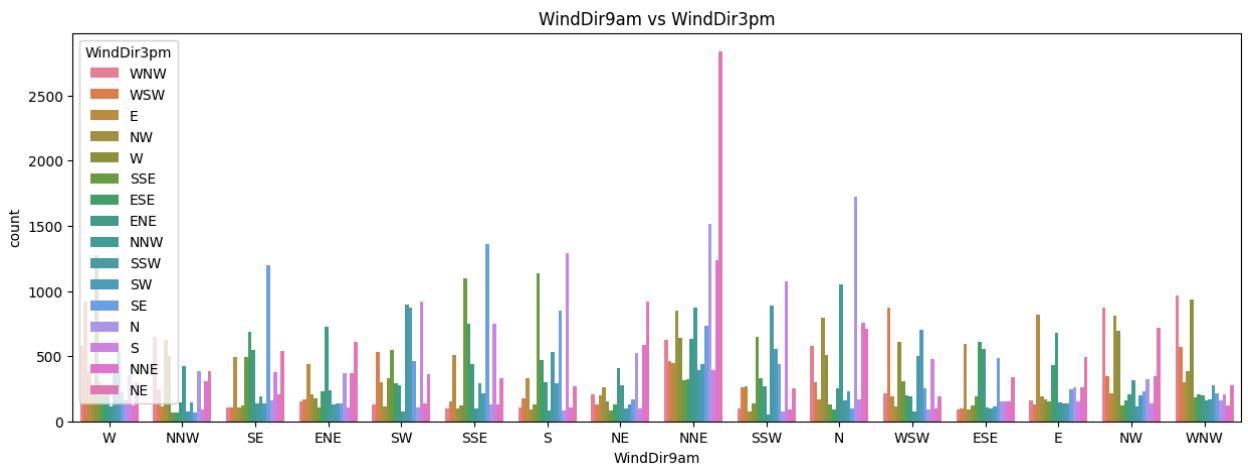
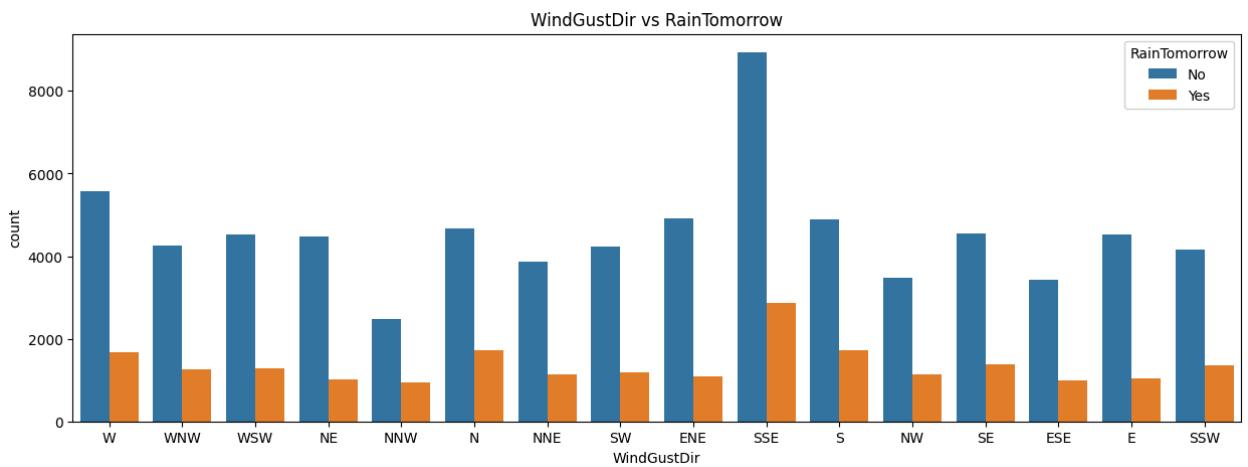
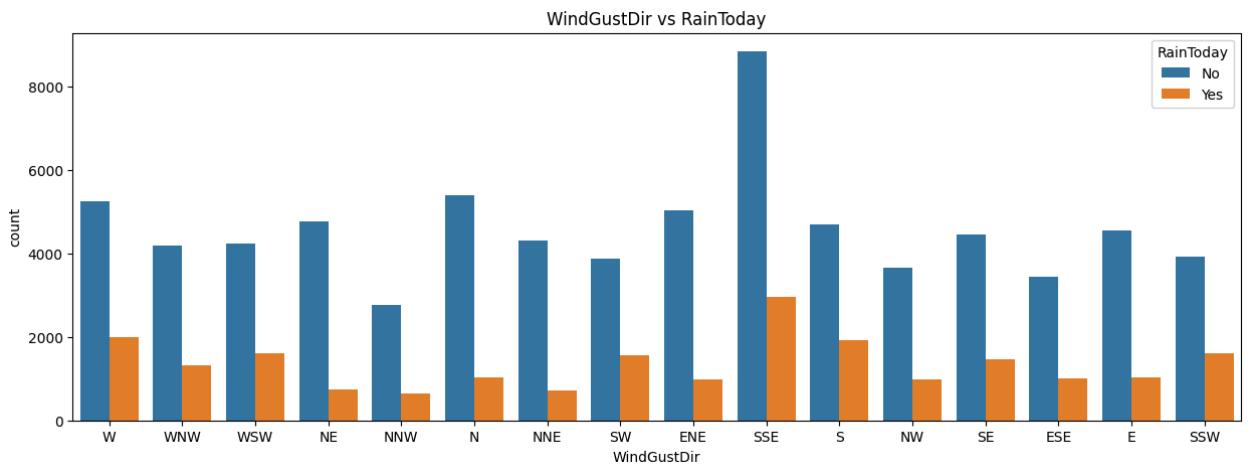


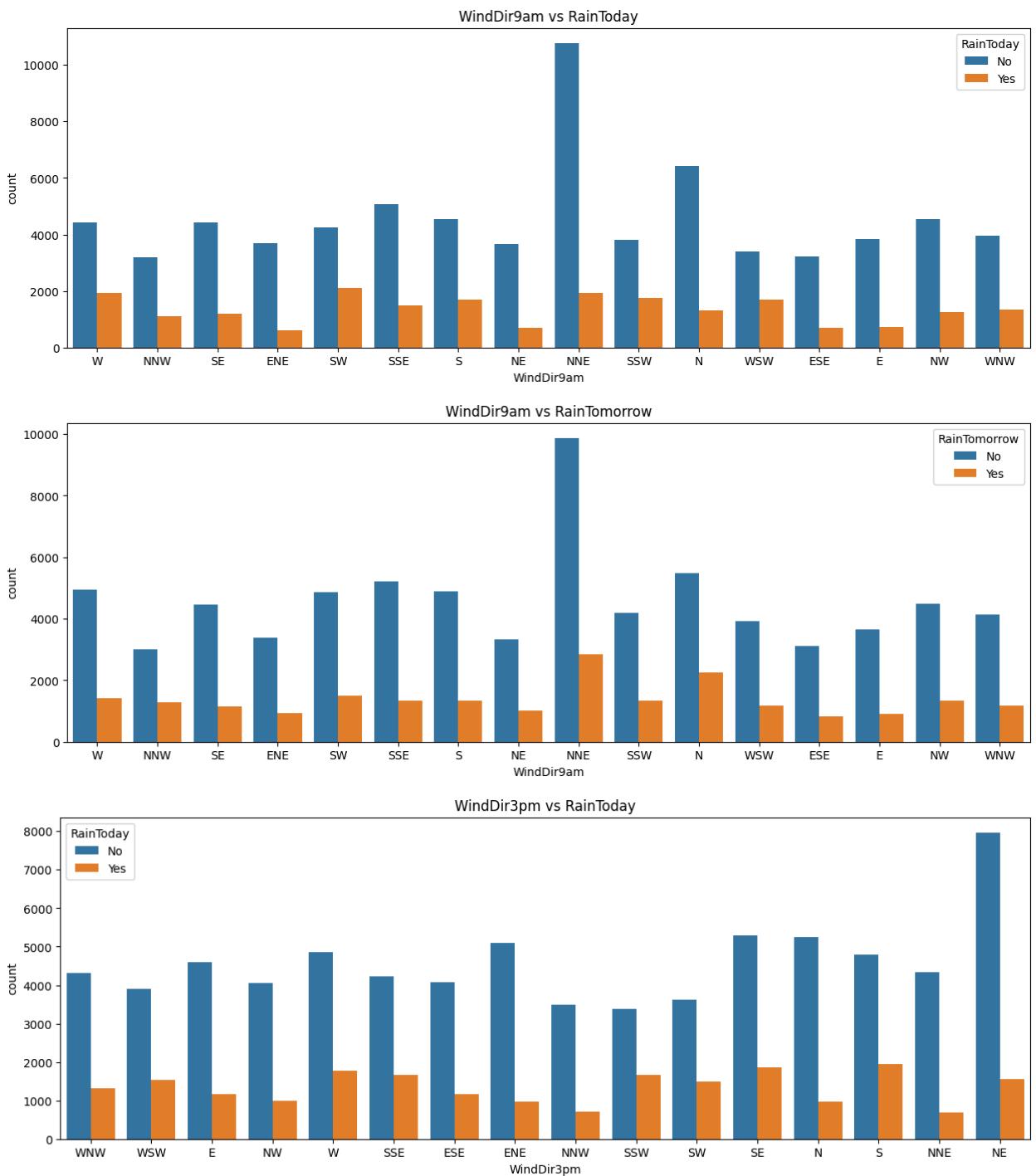


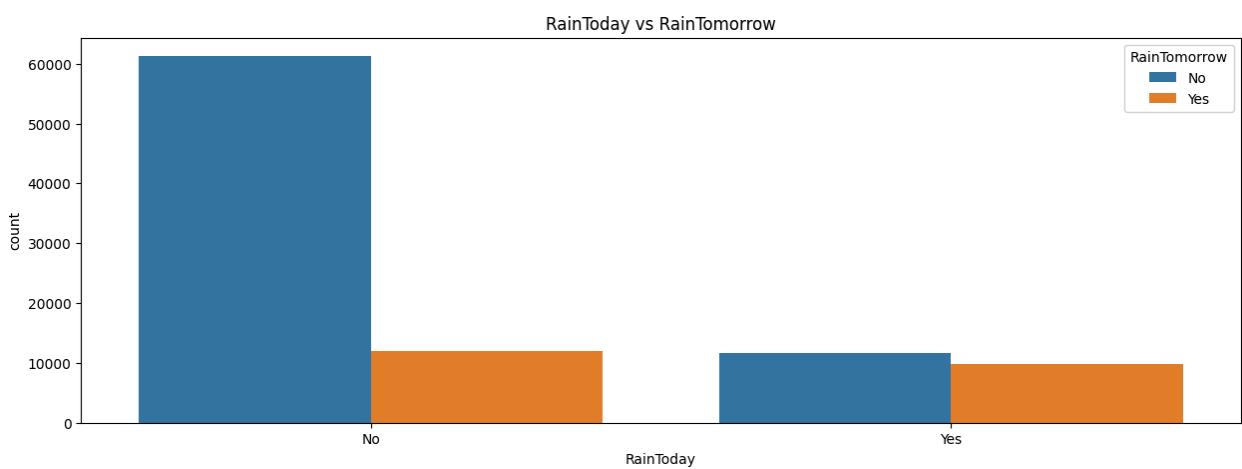
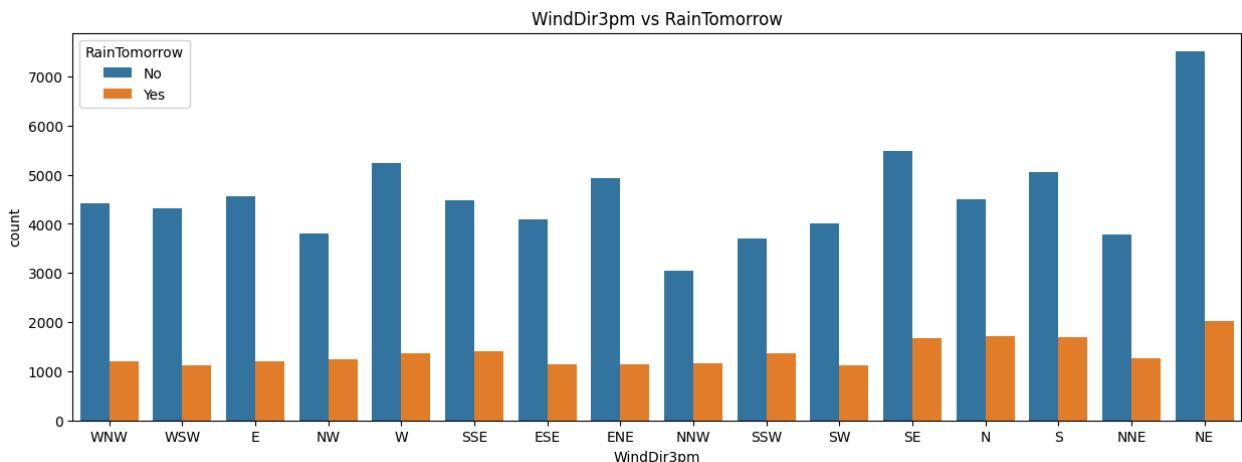


```
In [ ]: for i in range(len(categorical5.columns)):
    for j in range(i+1,len(categorical5.columns)):
        if categorical5.columns[i]!='Date' and categorical5.columns[j]!='Date' and
            plt.figure(figsize=(15,5))
            sns.countplot(x=df5[categorical5.columns[i]],hue=df5[categorical5.columns[j]])
            plt.title(categorical5.columns[i] +' vs '+categorical5.columns[j])
            plt.show()
```

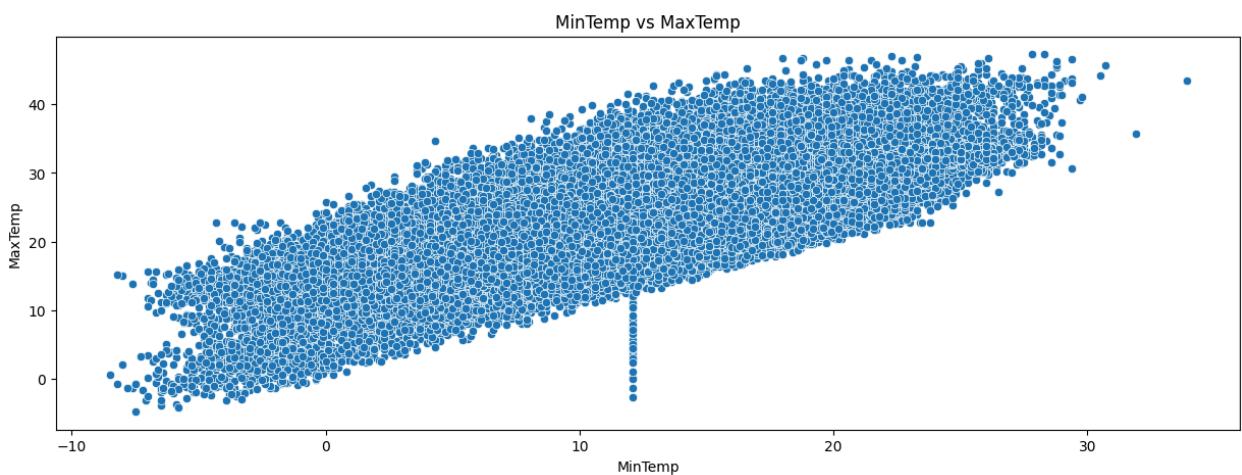


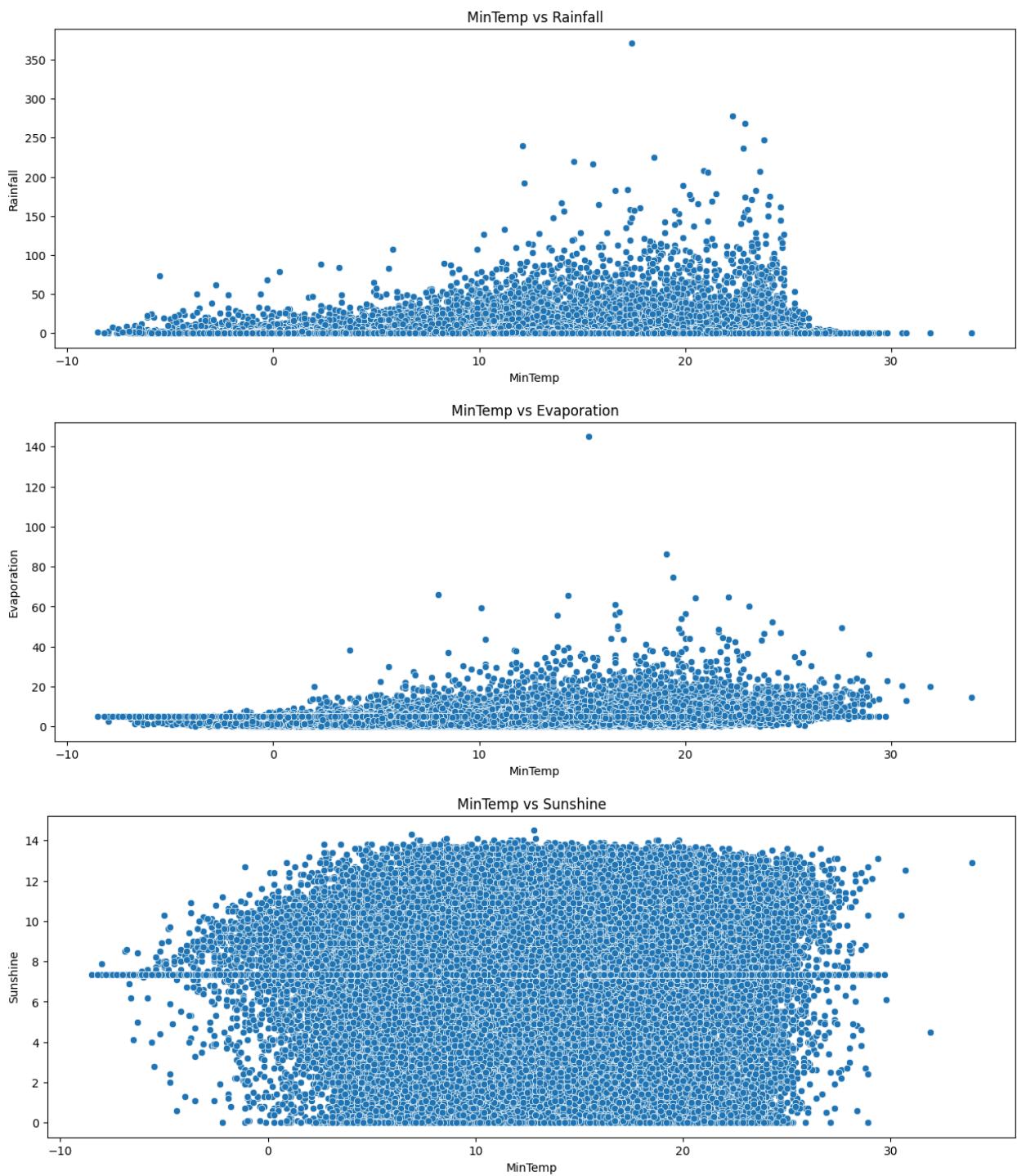


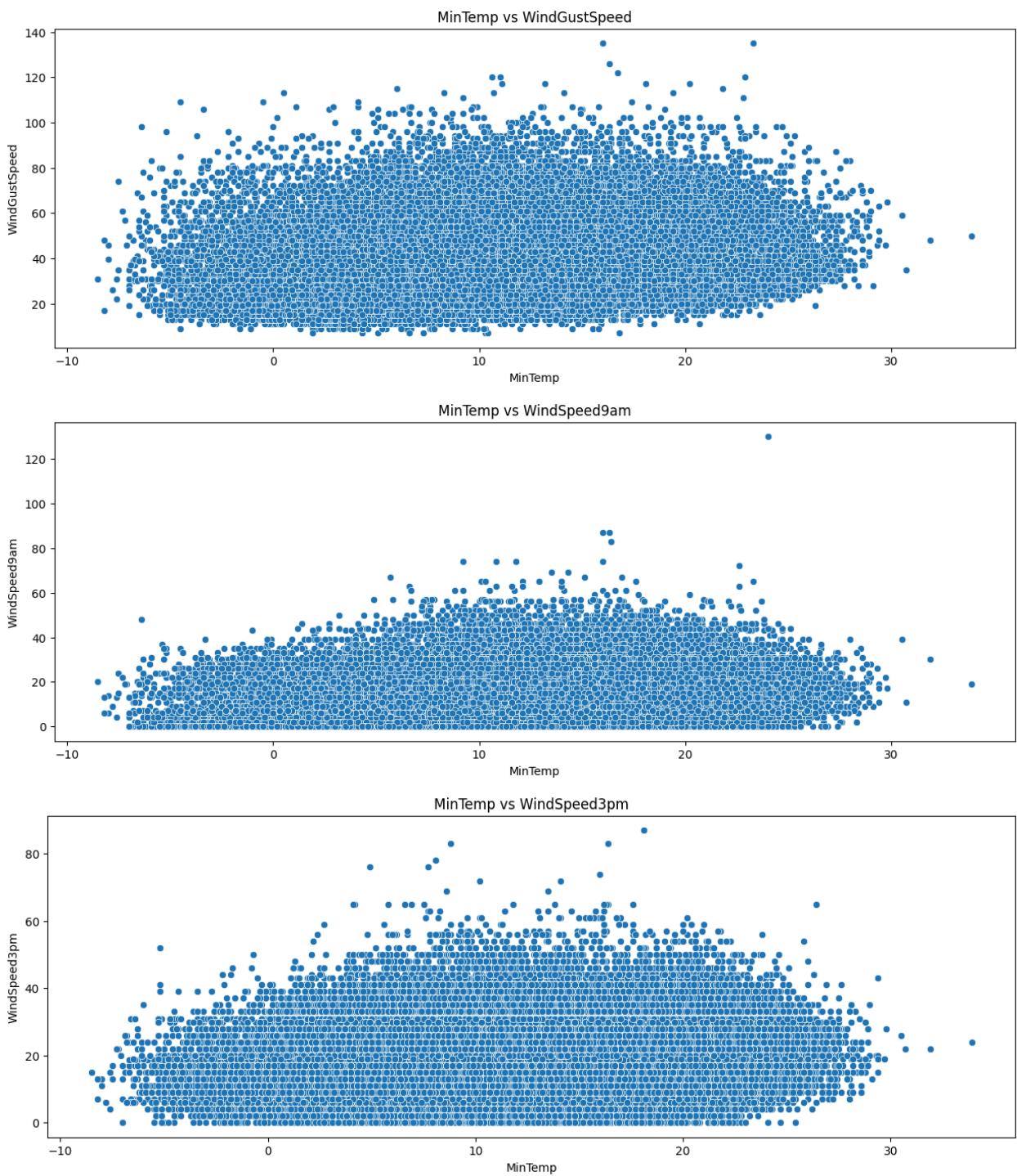


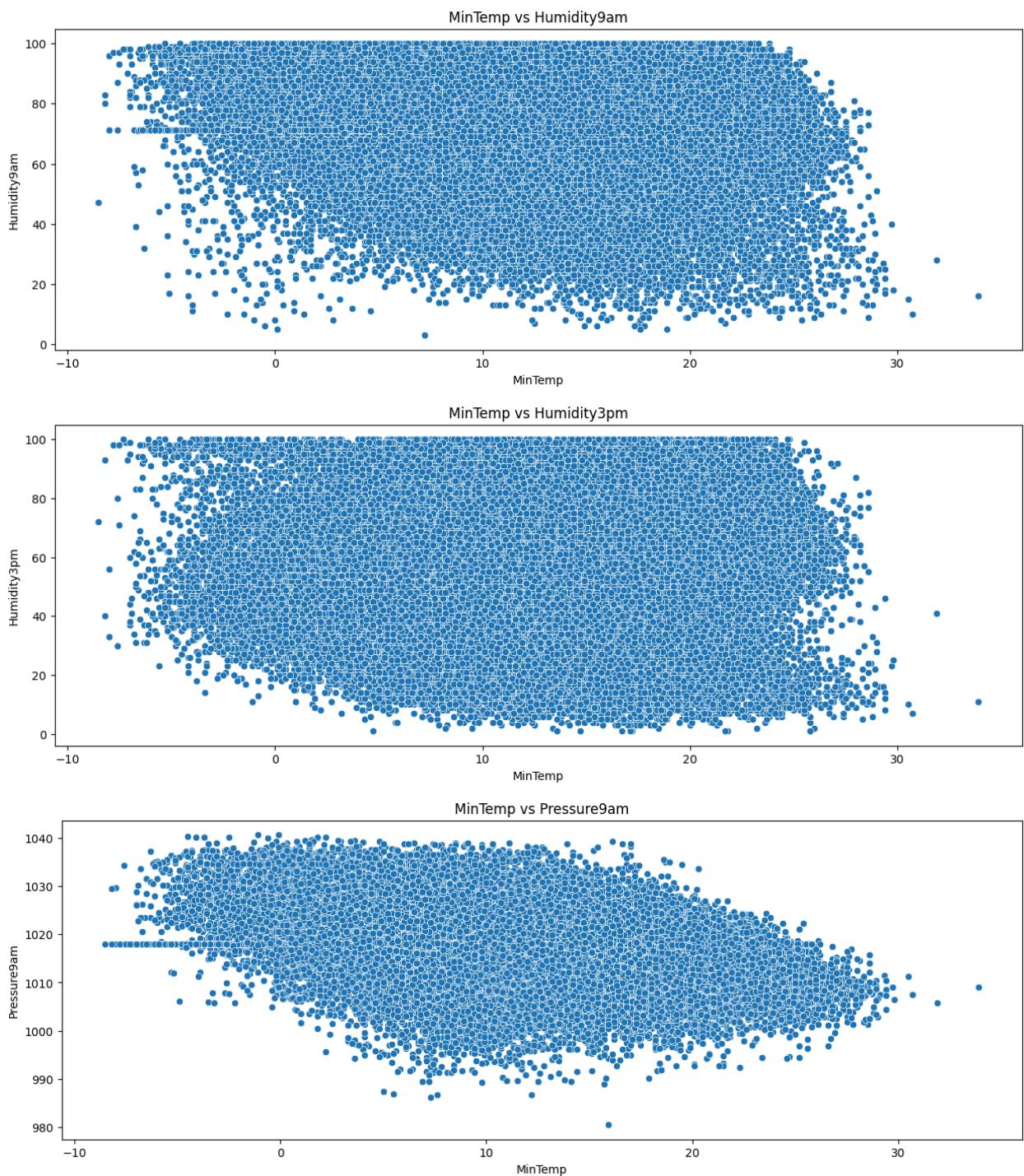


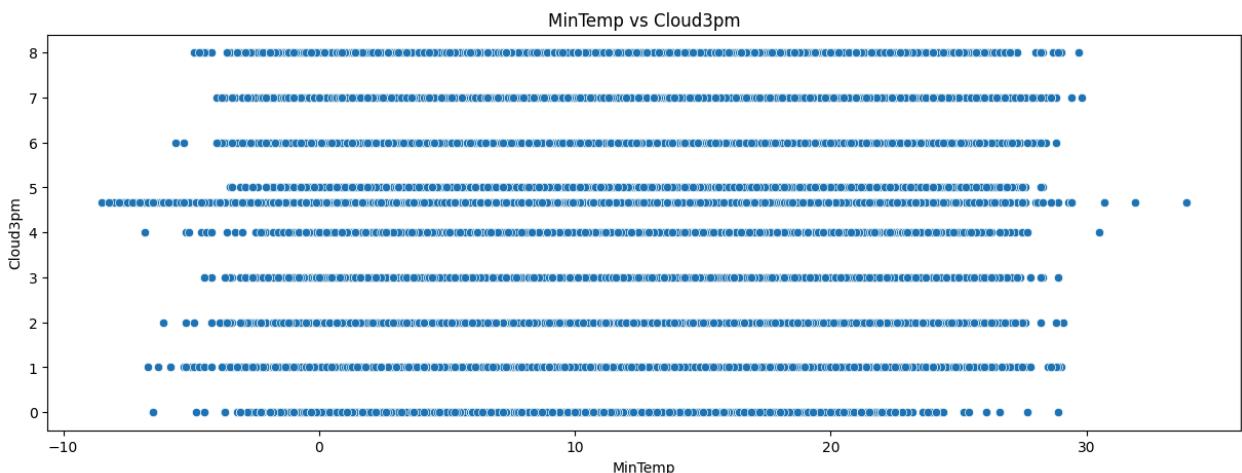
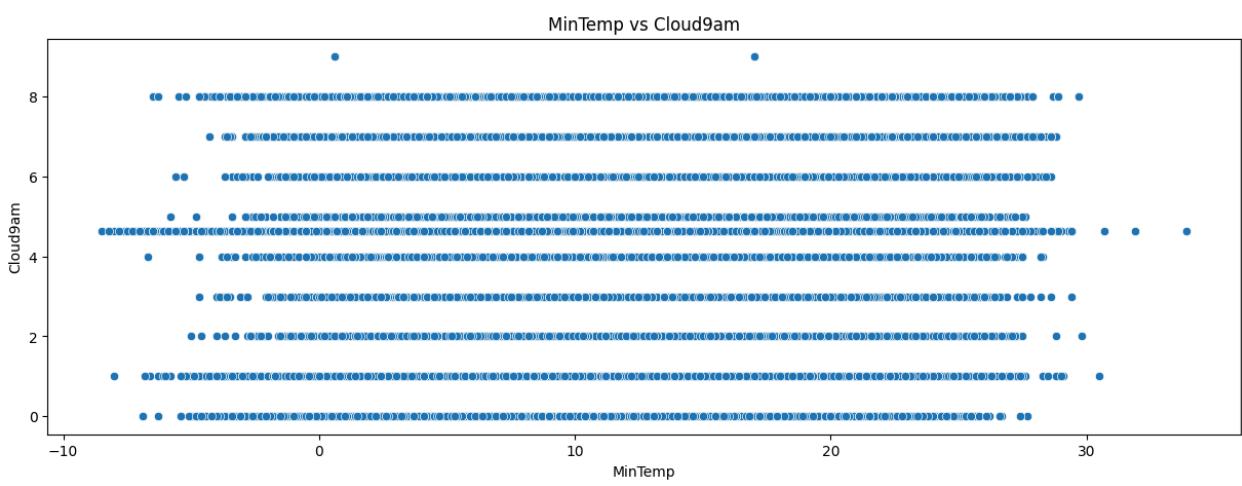
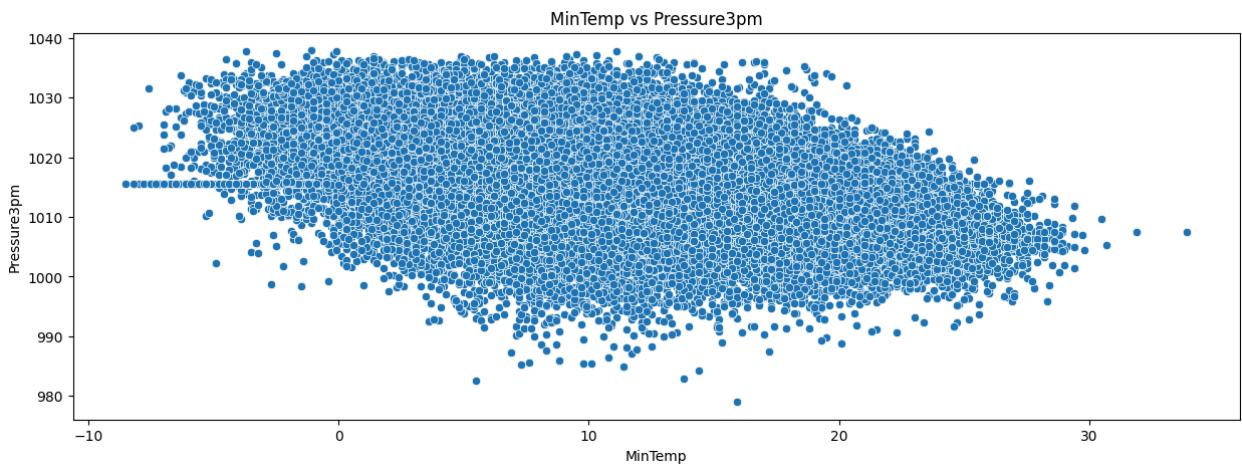
```
In [ ]: for i in range(len(Numerical5.columns)):
    for j in range(i+1,len(Numerical5.columns)):
        plt.figure(figsize=(15,5))
        sns.scatterplot(x=df5[Numerical5.columns[i]],y=df5[Numerical5.columns[j]])
        plt.title(Numerical5.columns[i] + ' vs ' + Numerical5.columns[j])
        plt.show()
```

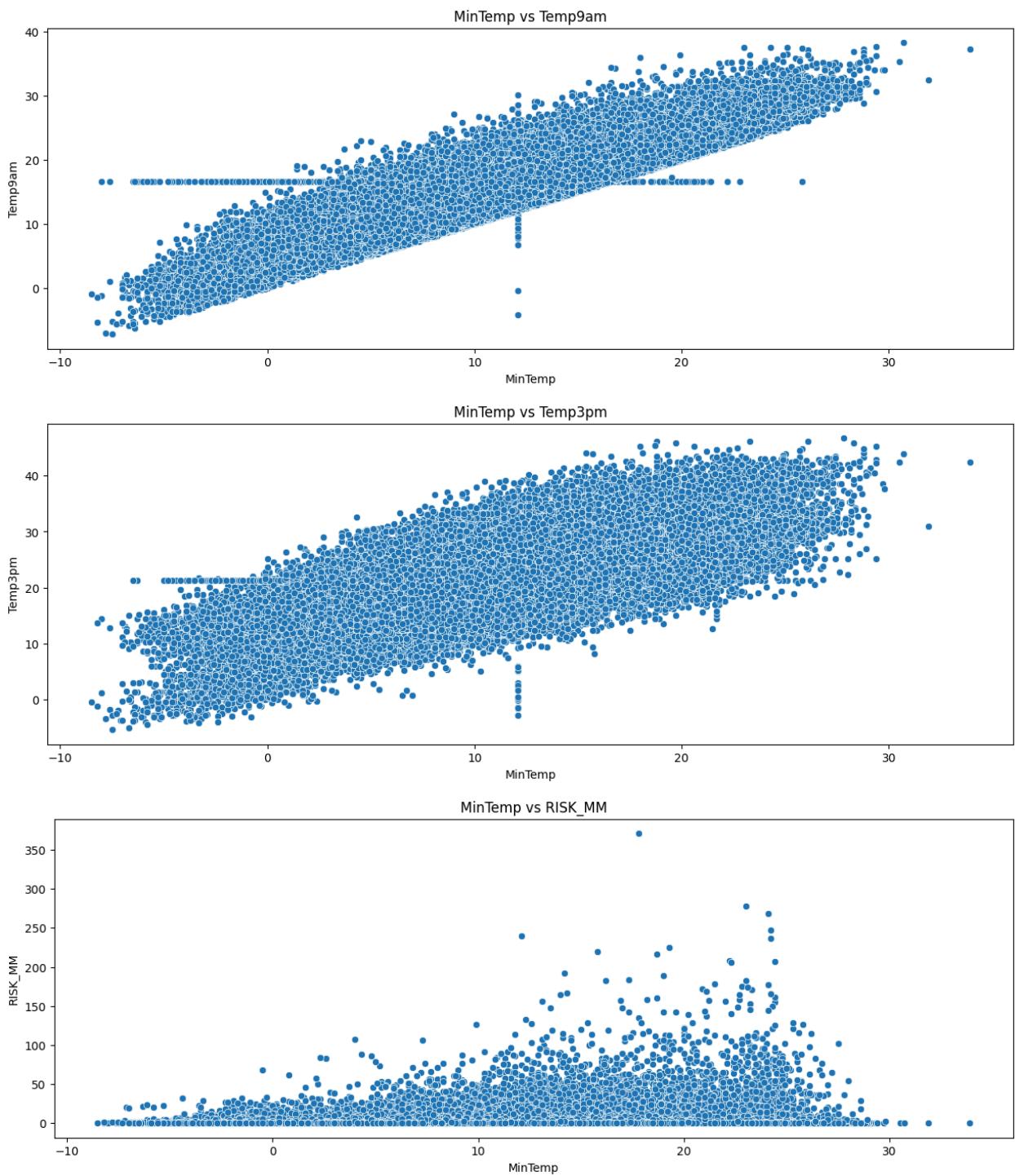




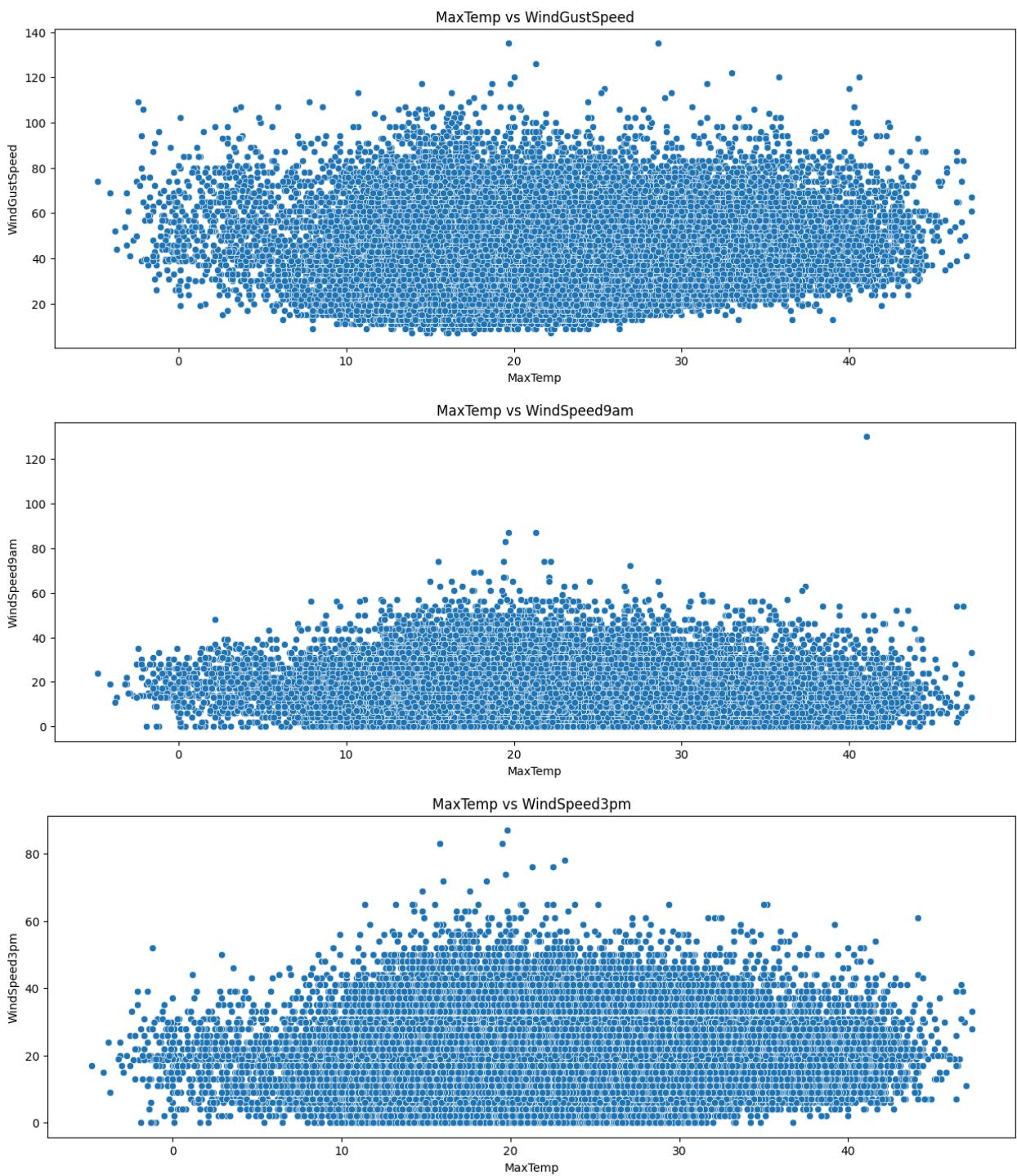


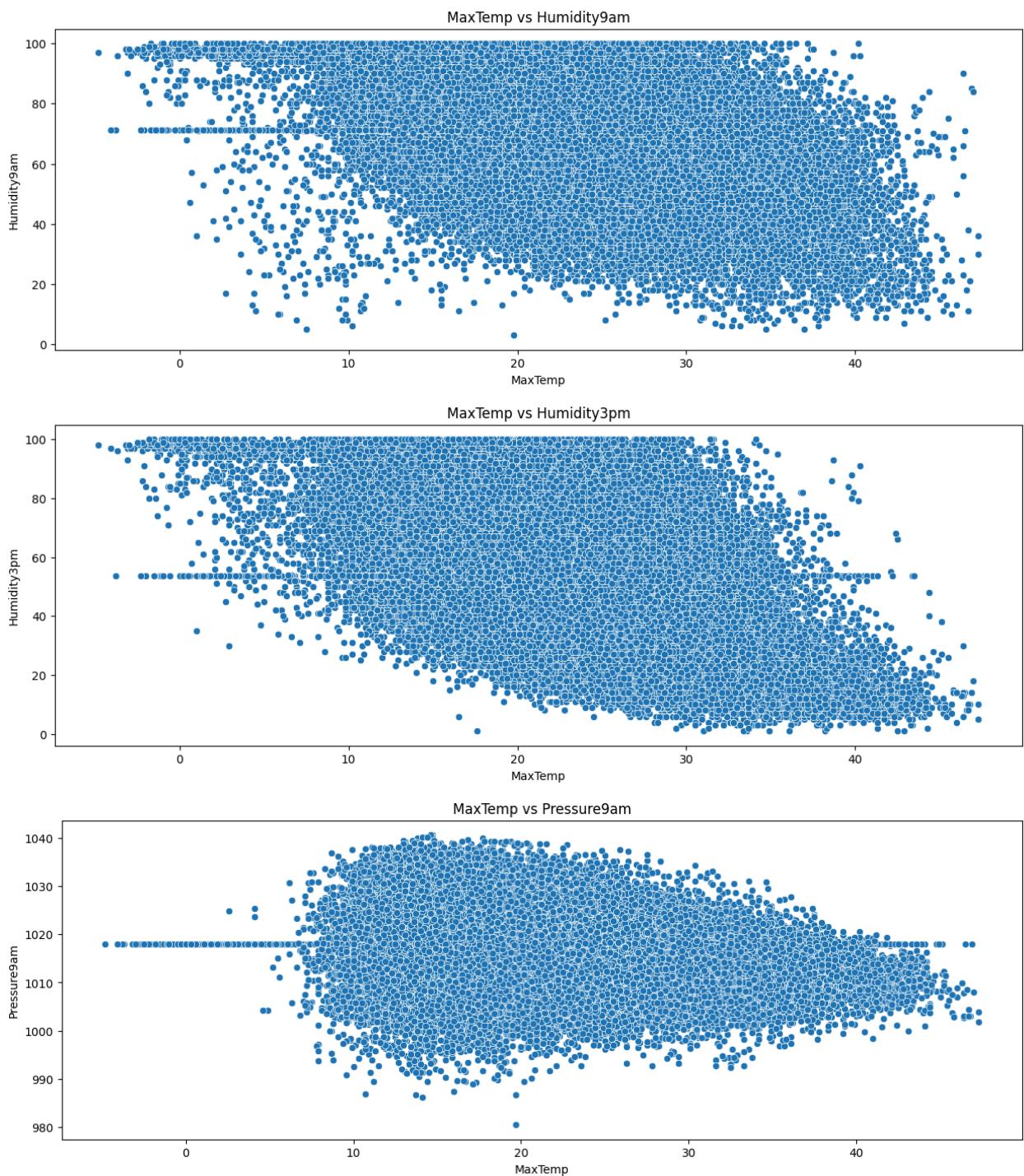


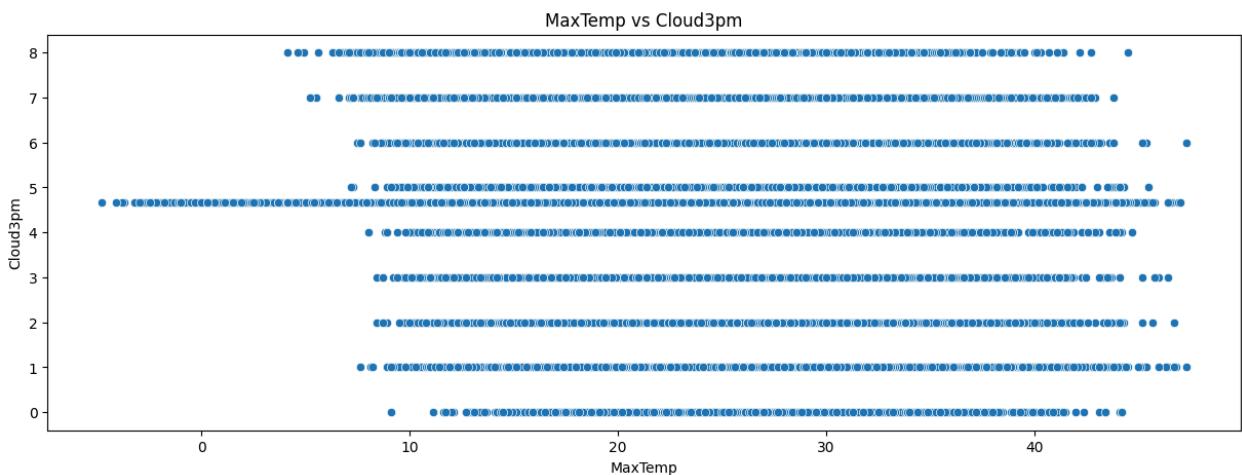
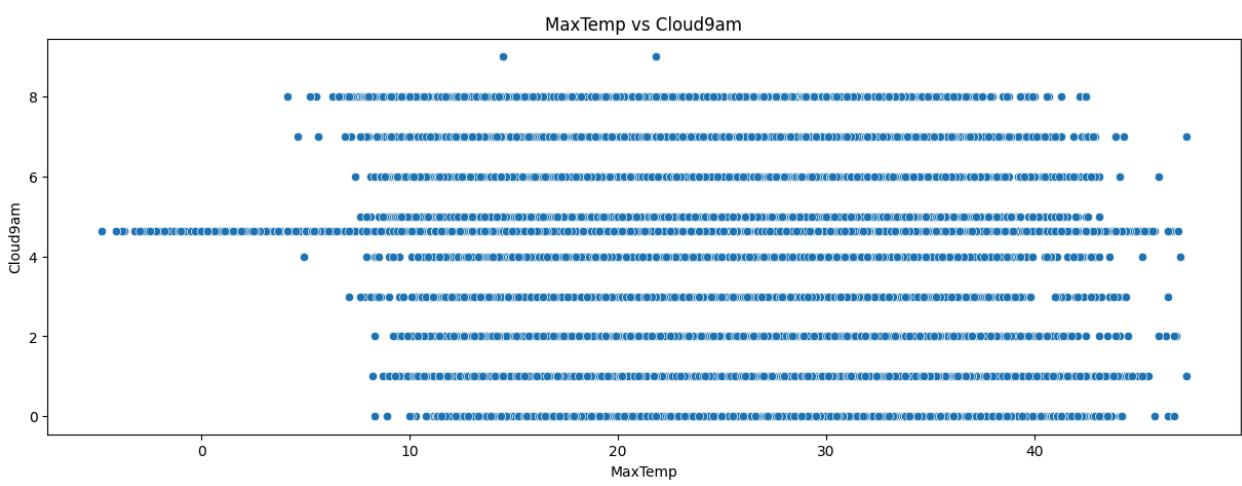
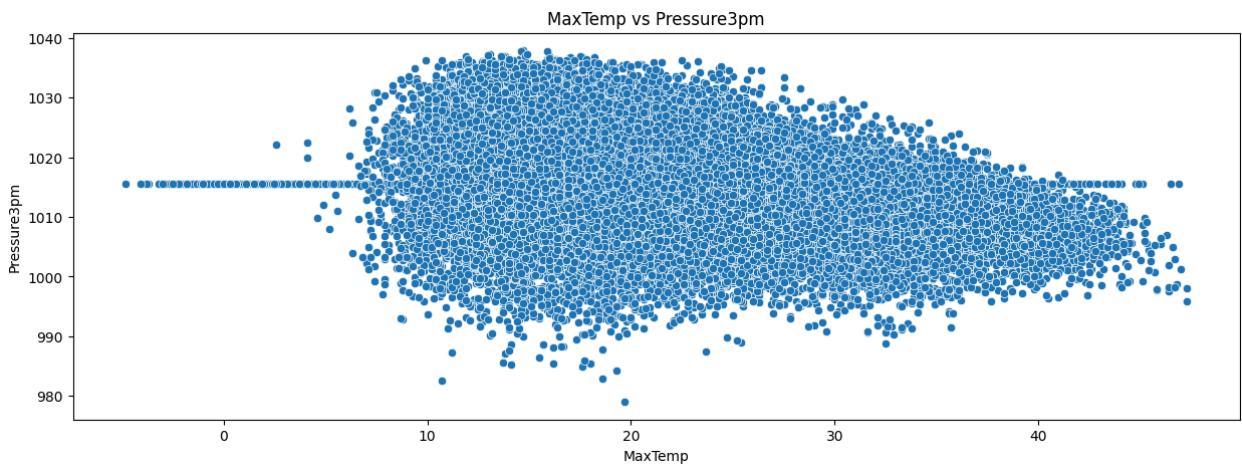


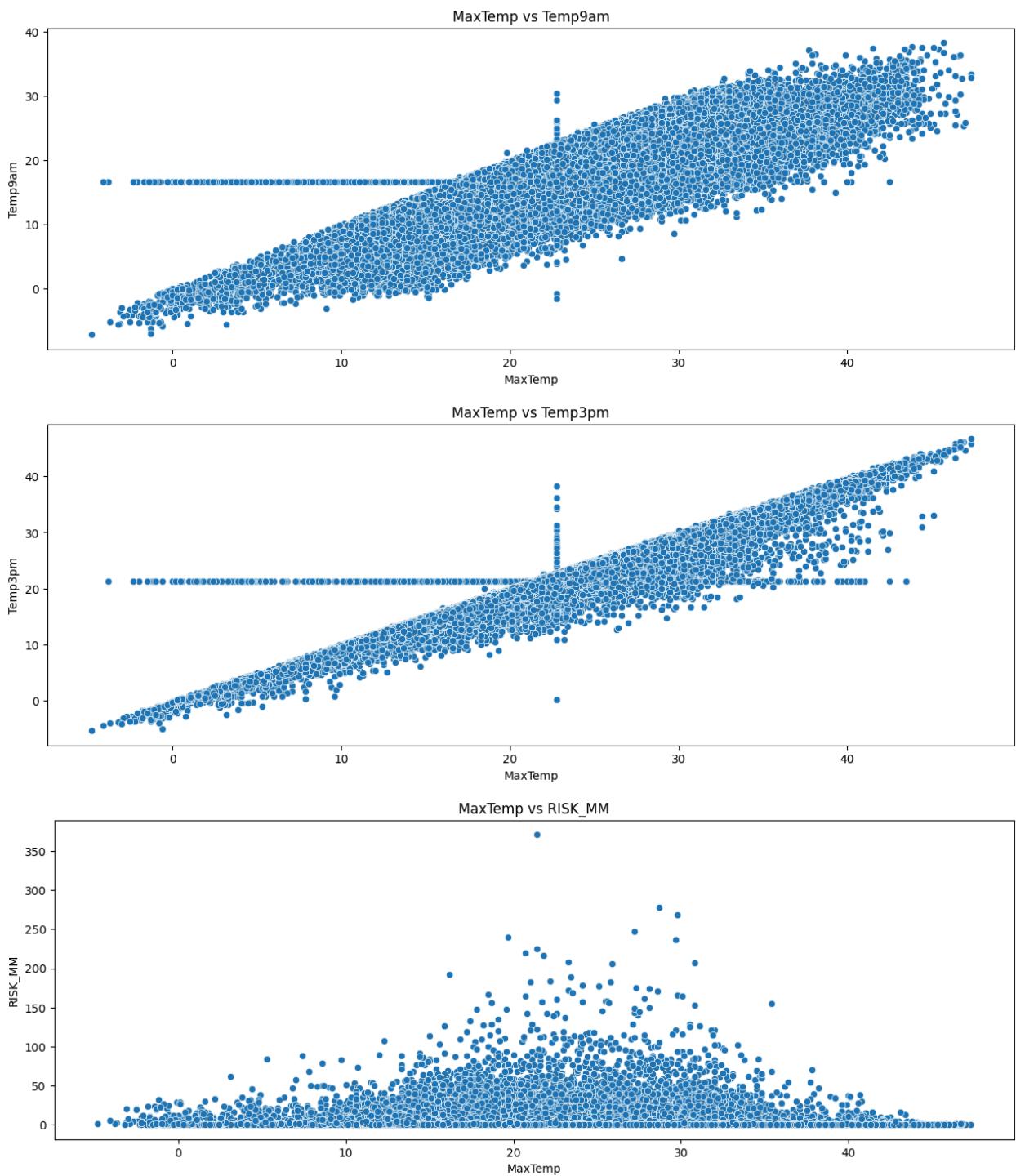


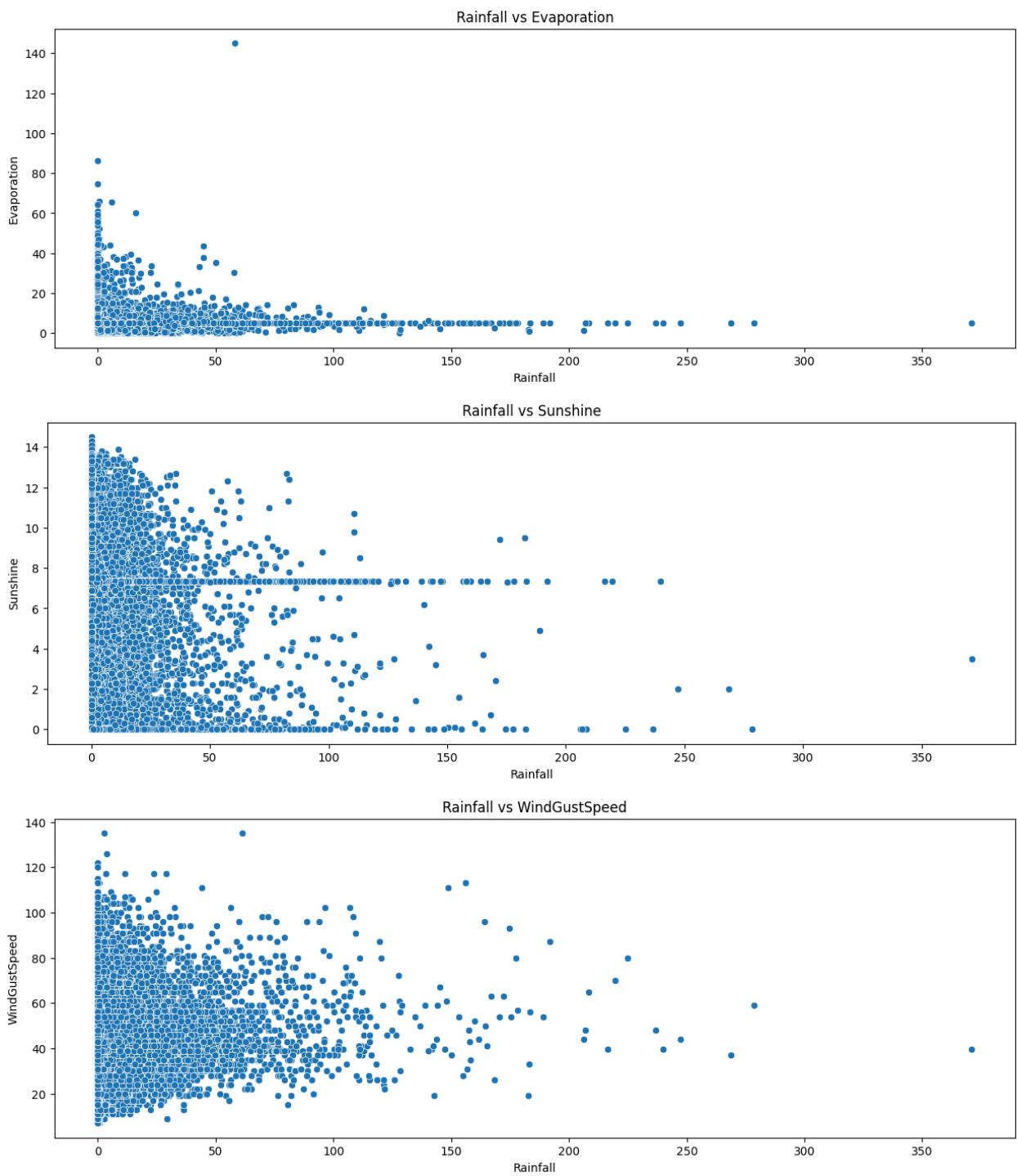




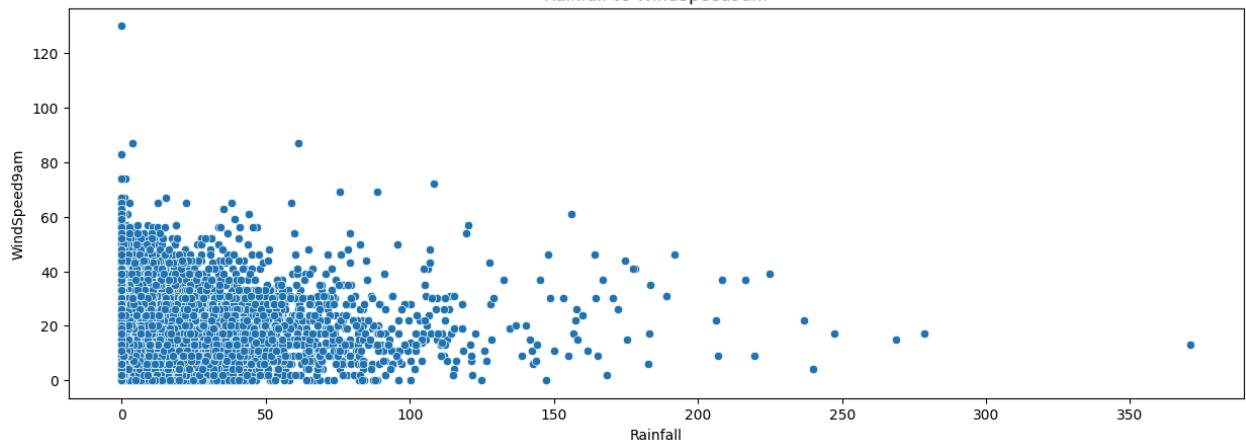




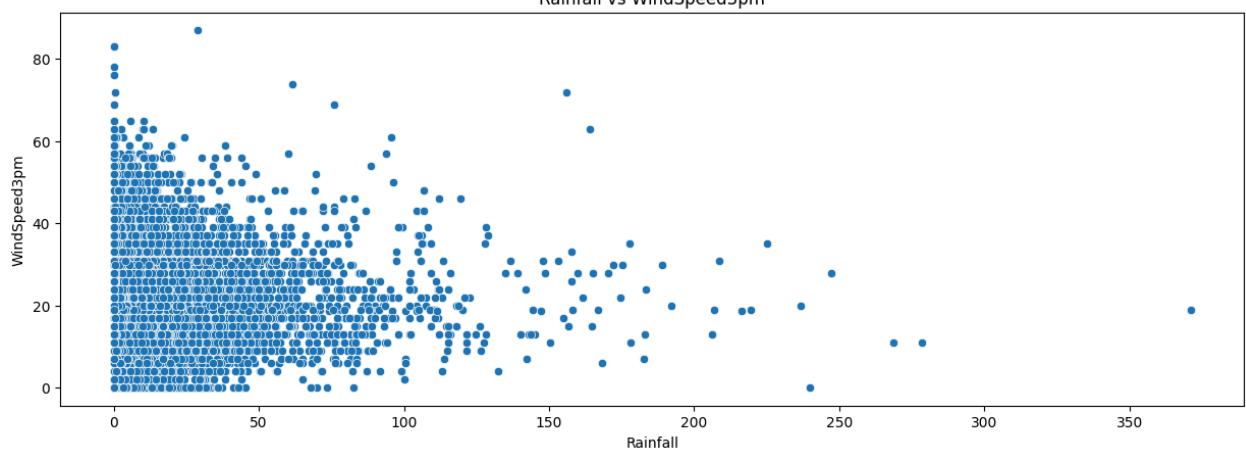




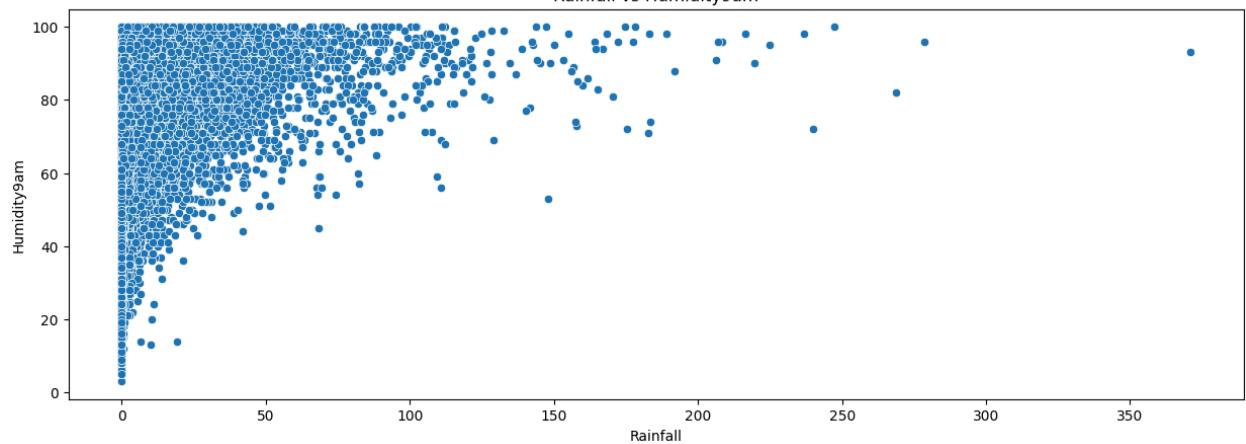
Rainfall vs WindSpeed9am

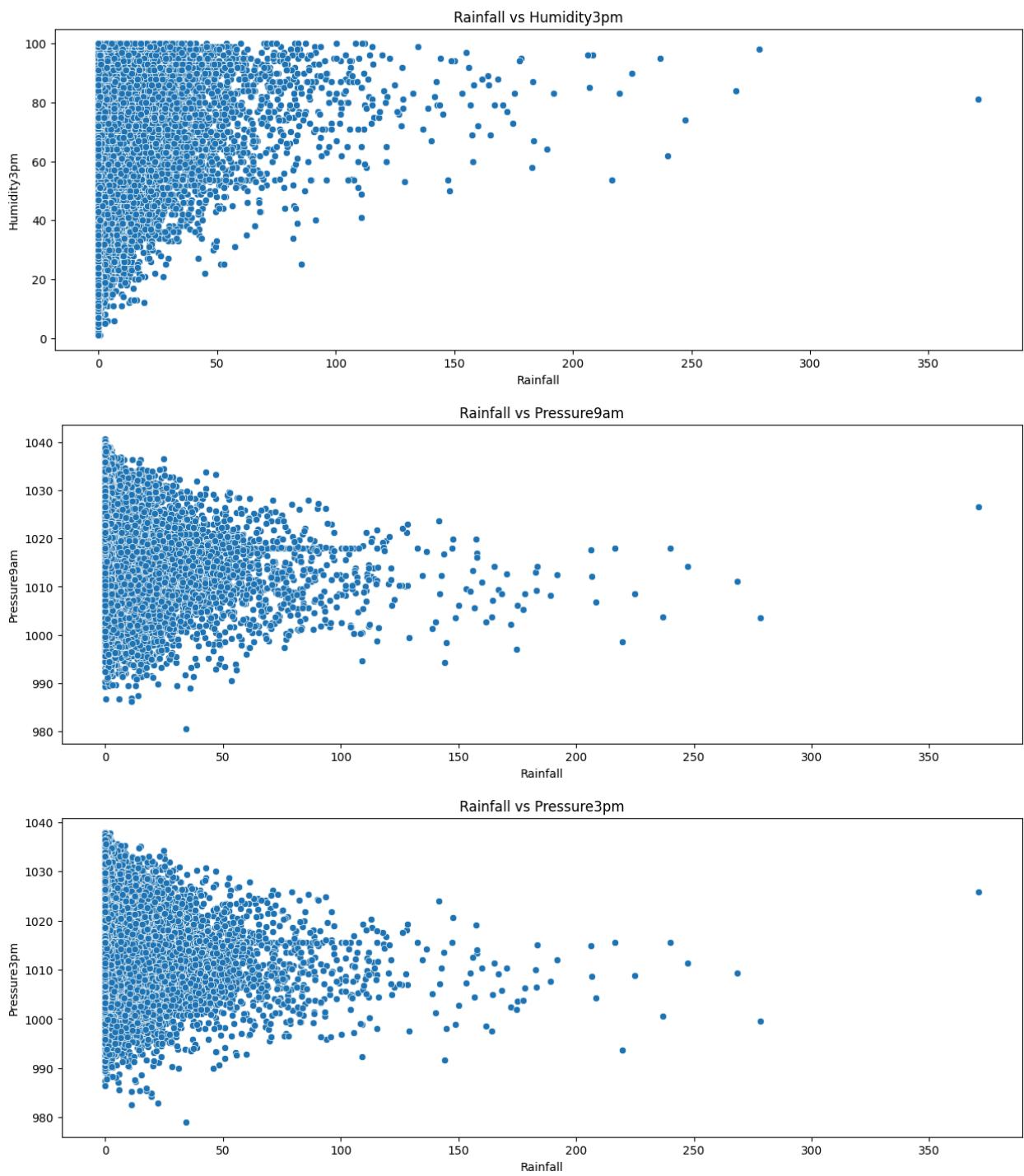


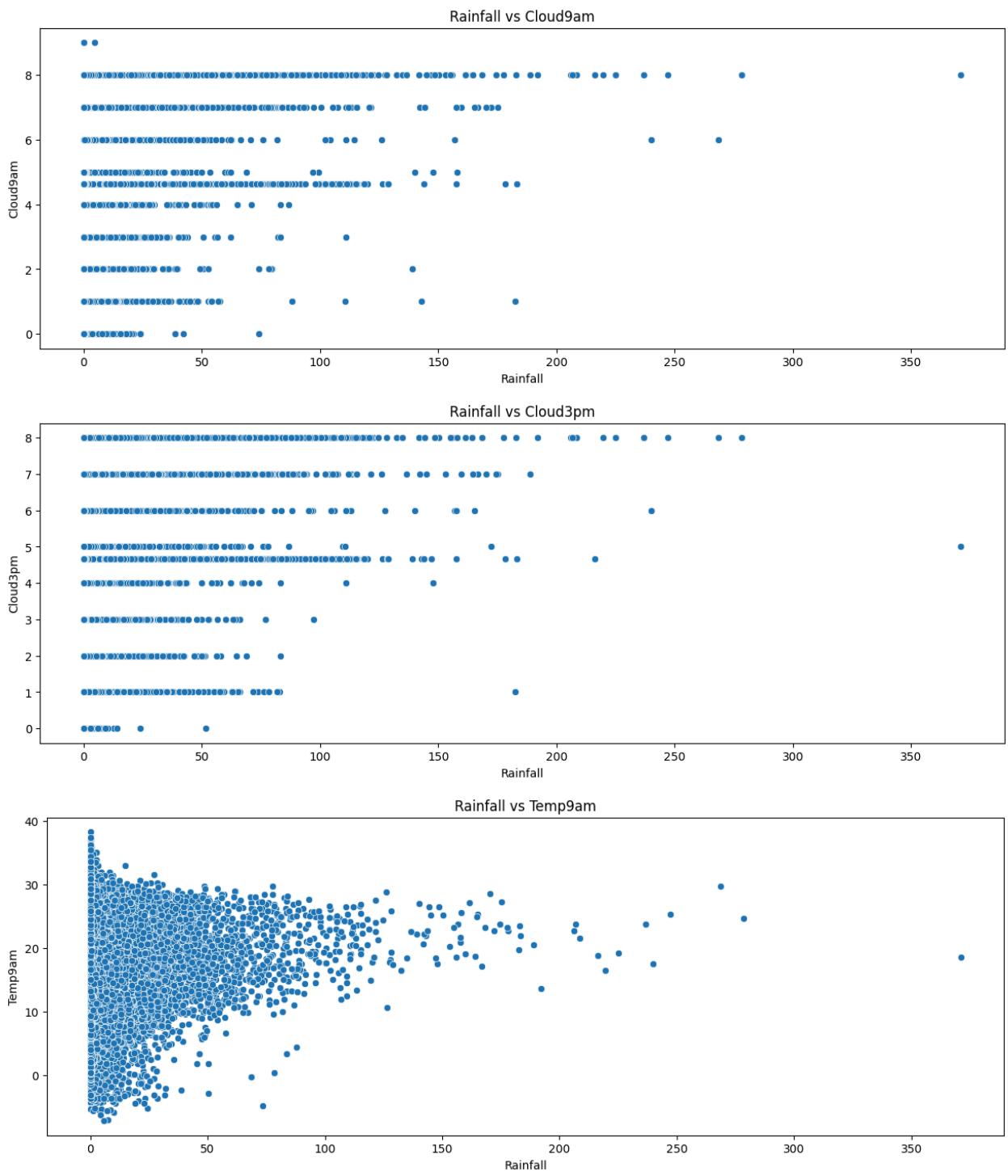
Rainfall vs WindSpeed3pm



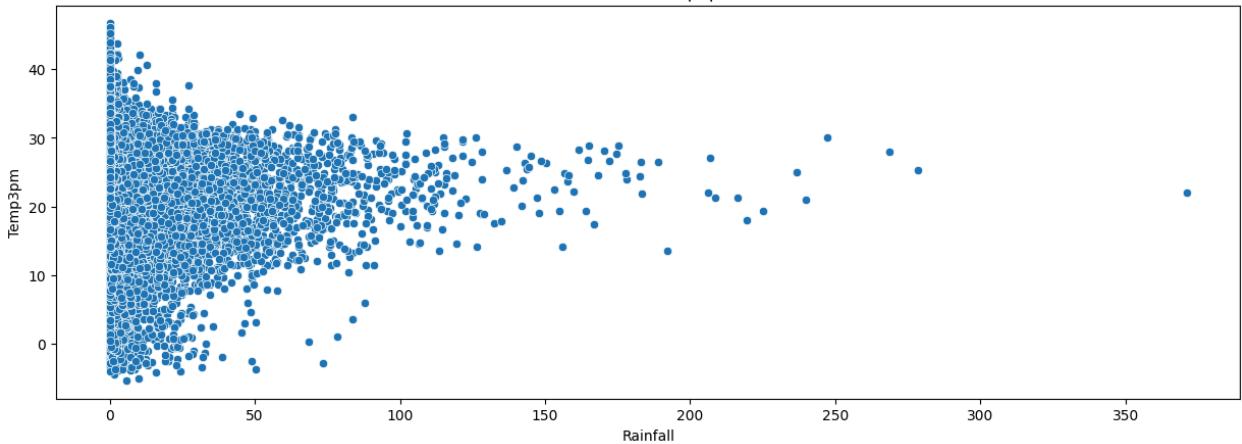
Rainfall vs Humidity9am



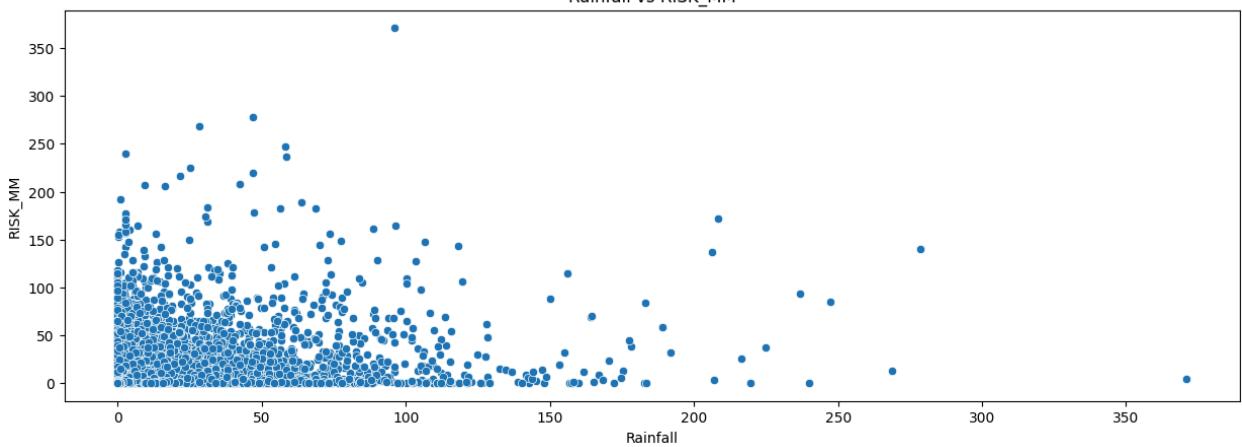




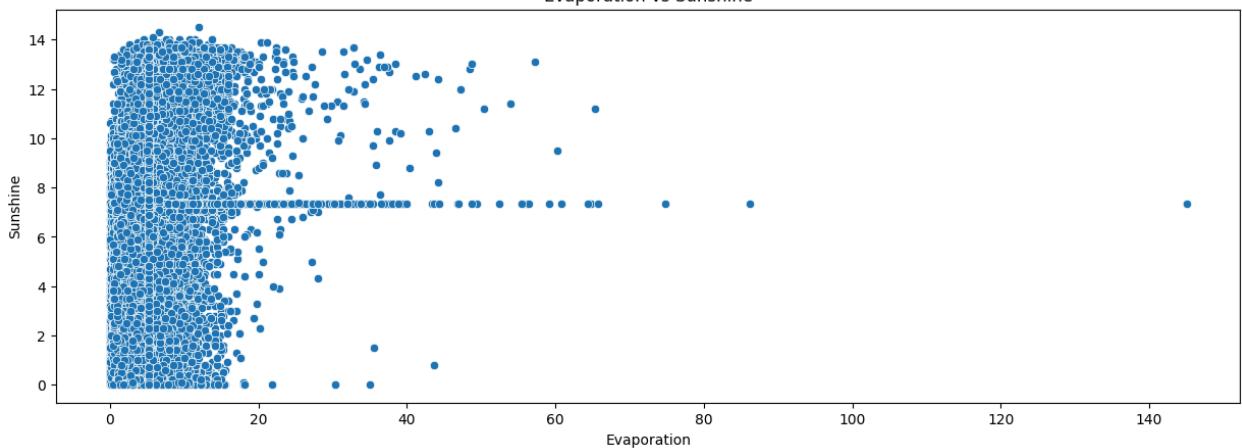
Rainfall vs Temp3pm

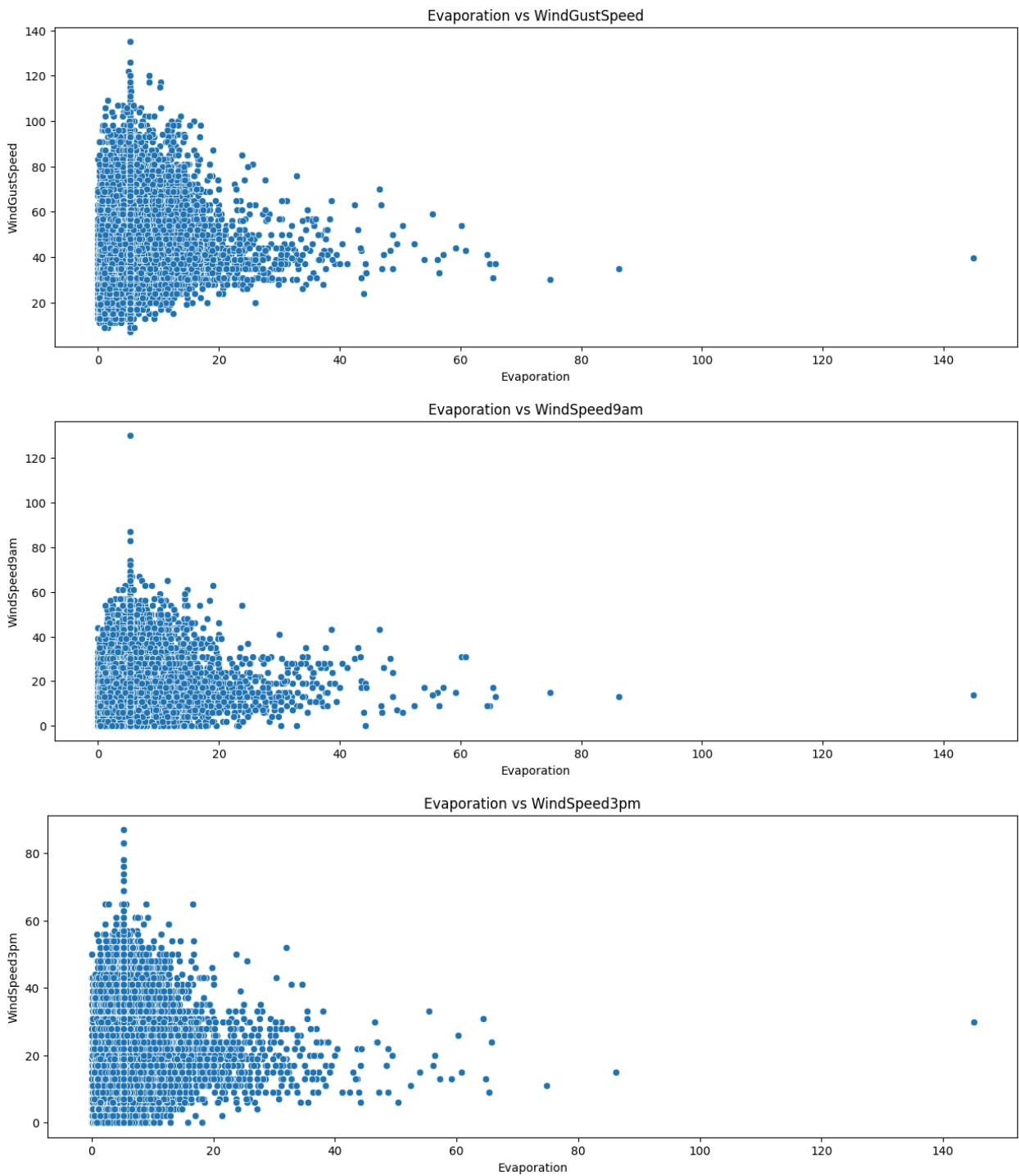


Rainfall vs RISK\_MM

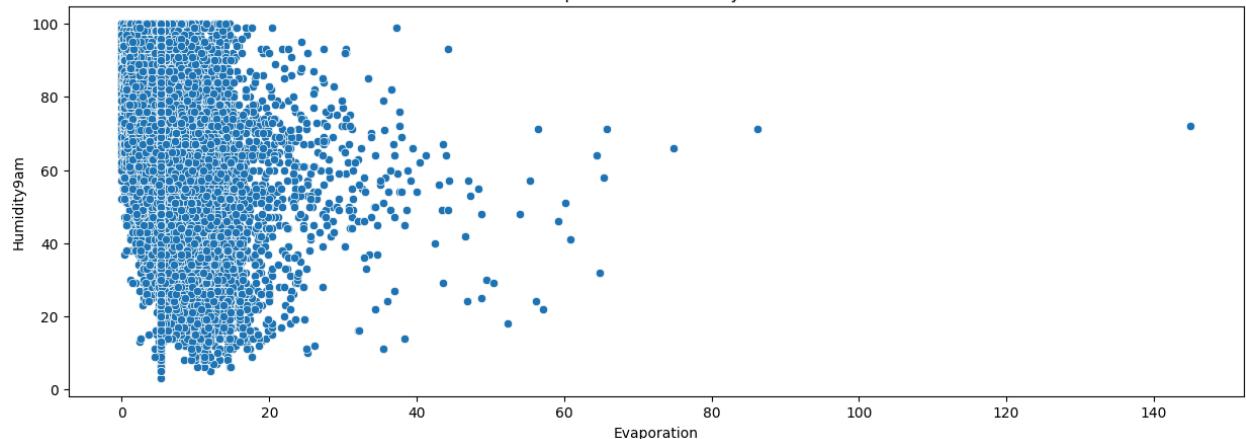


Evaporation vs Sunshine

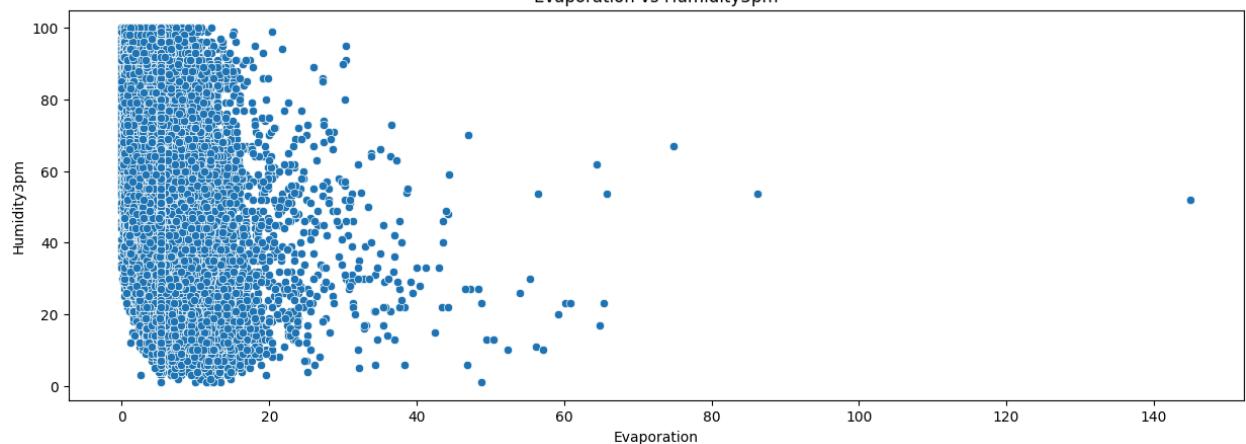




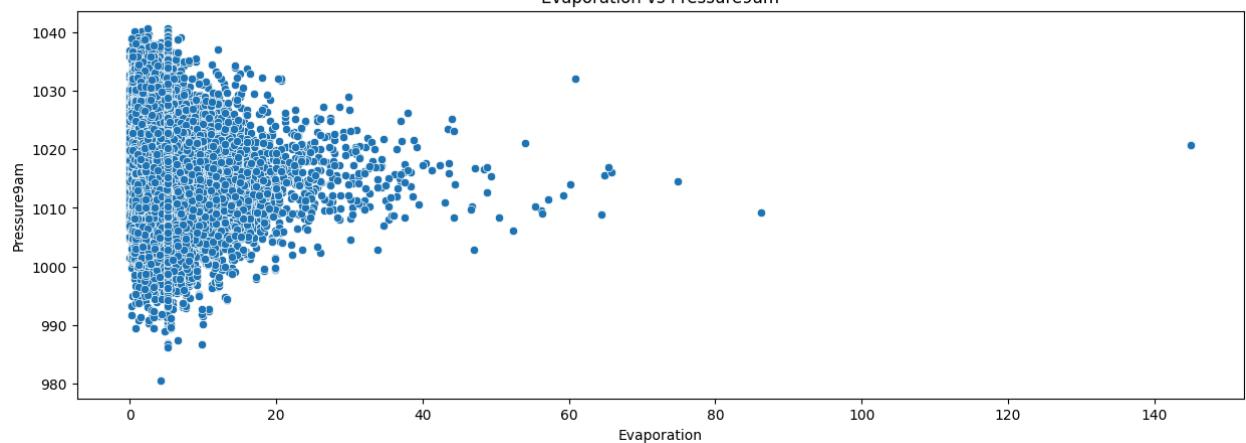
Evaporation vs Humidity9am

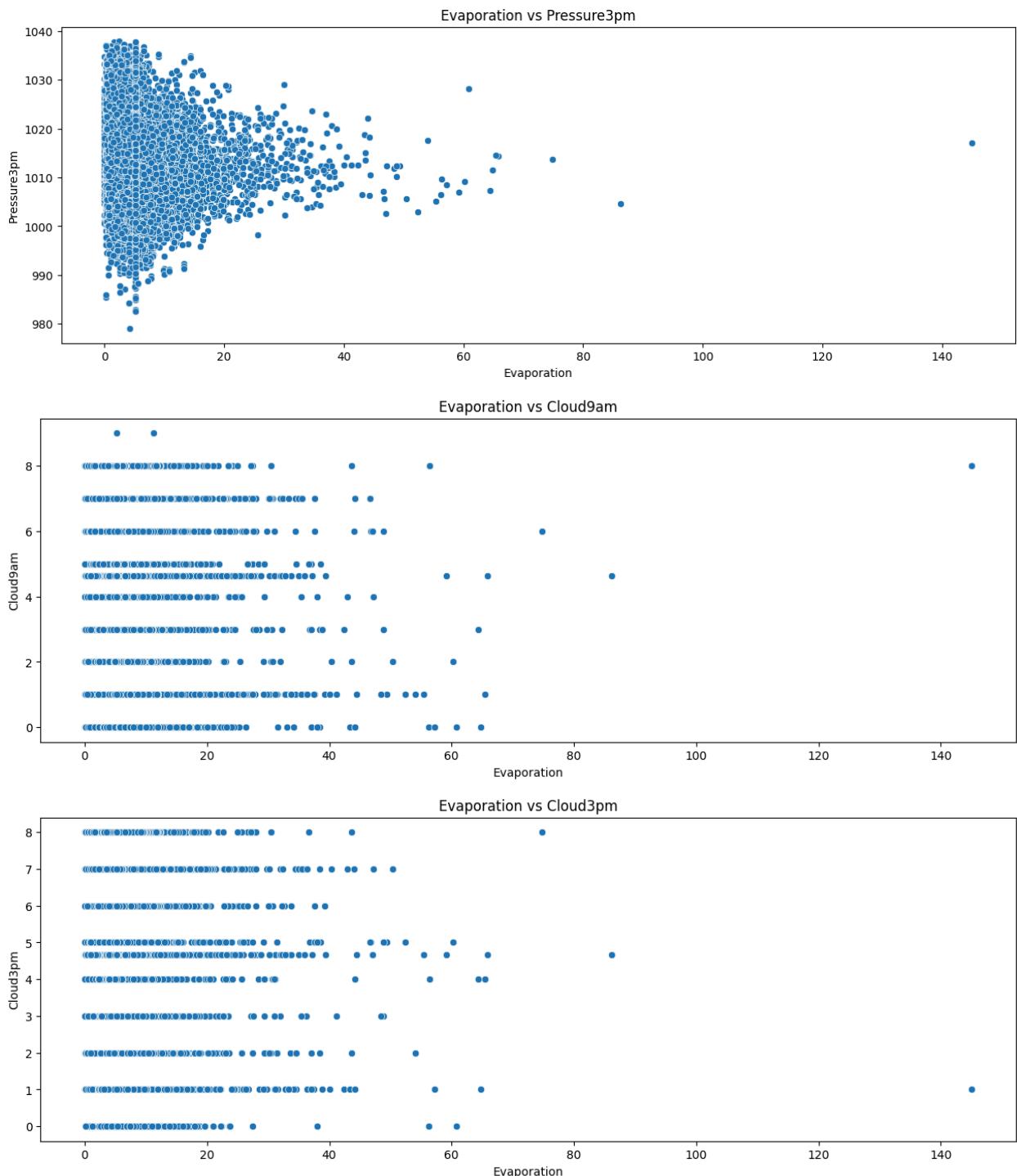


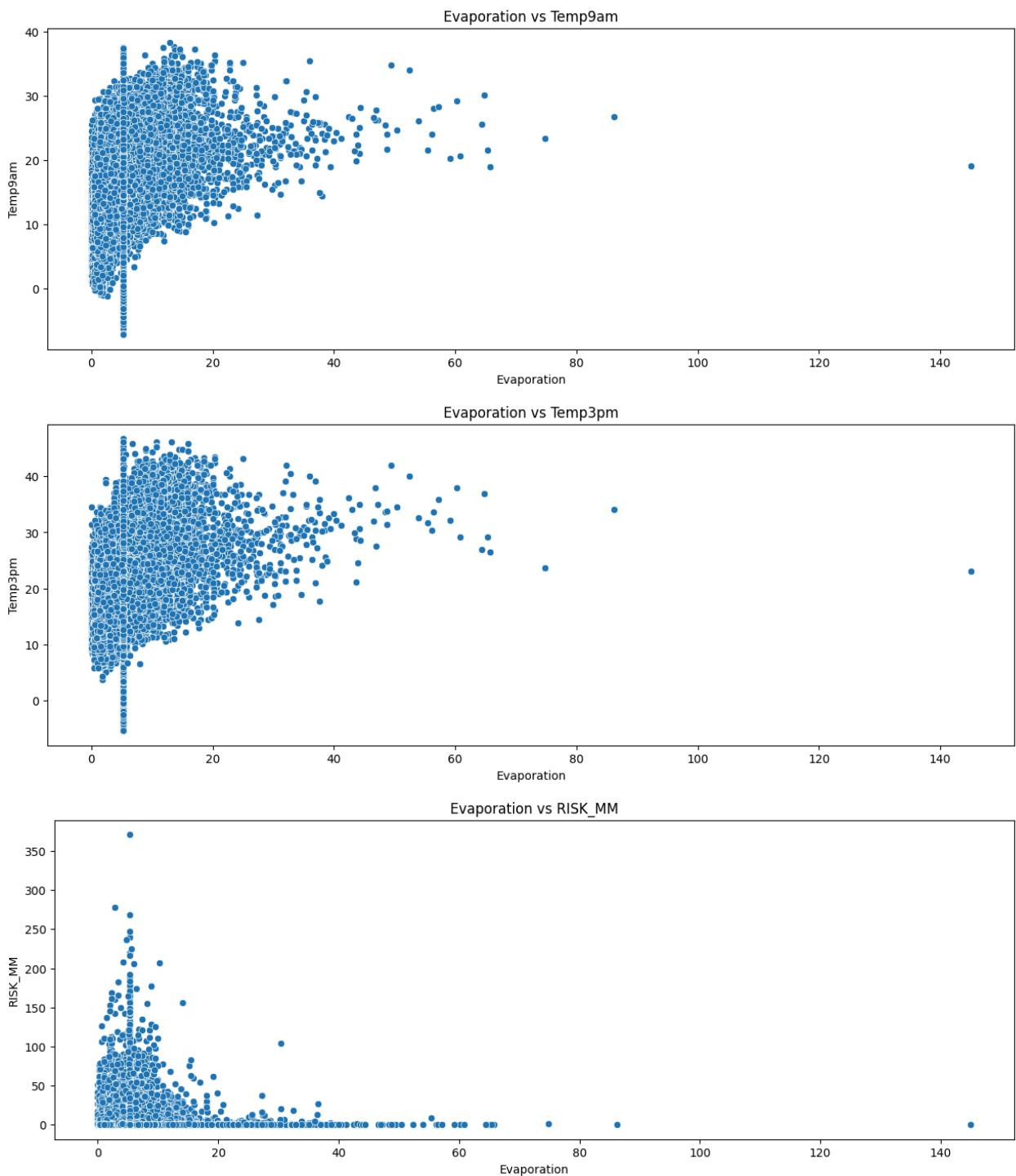
Evaporation vs Humidity3pm

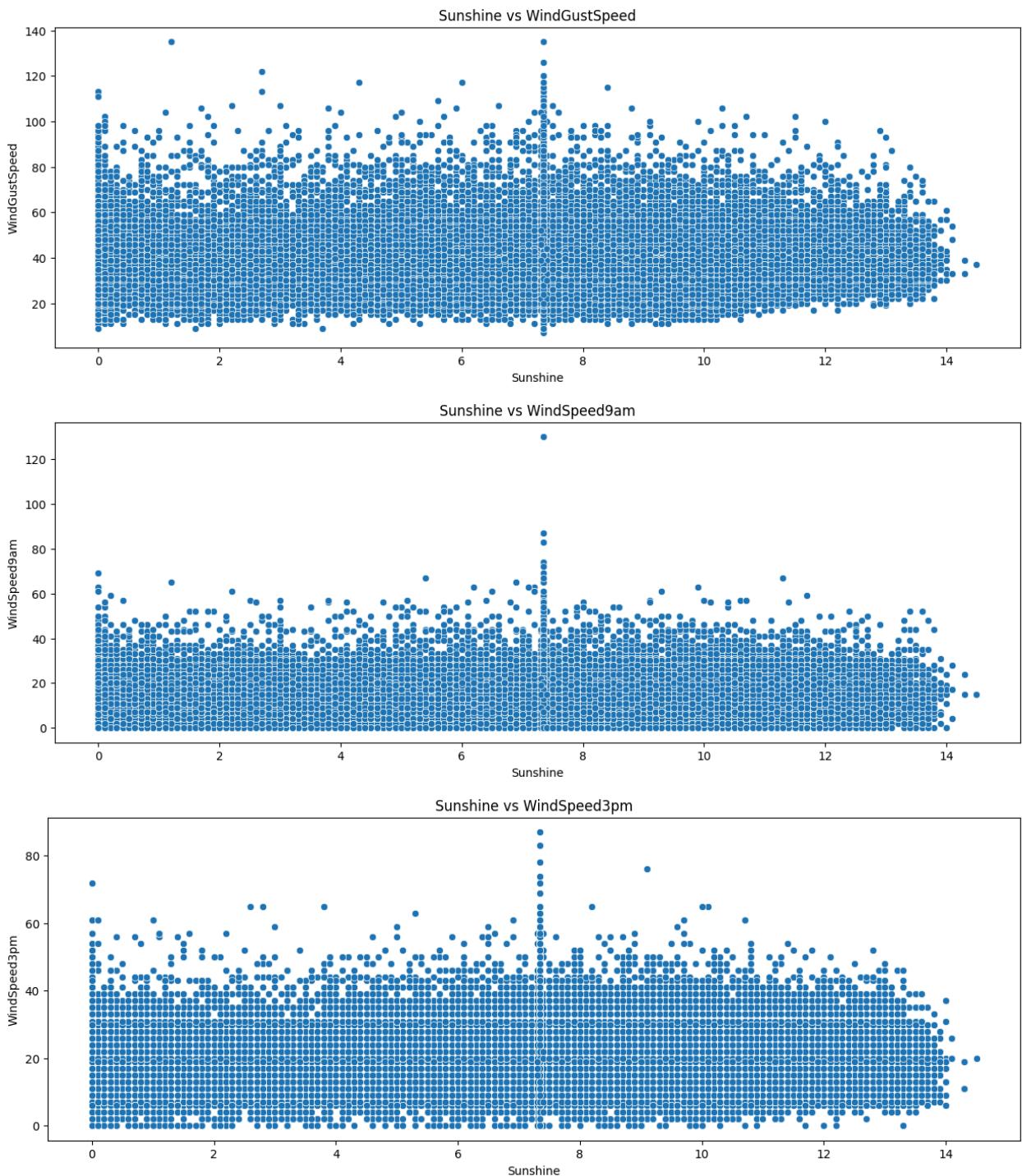


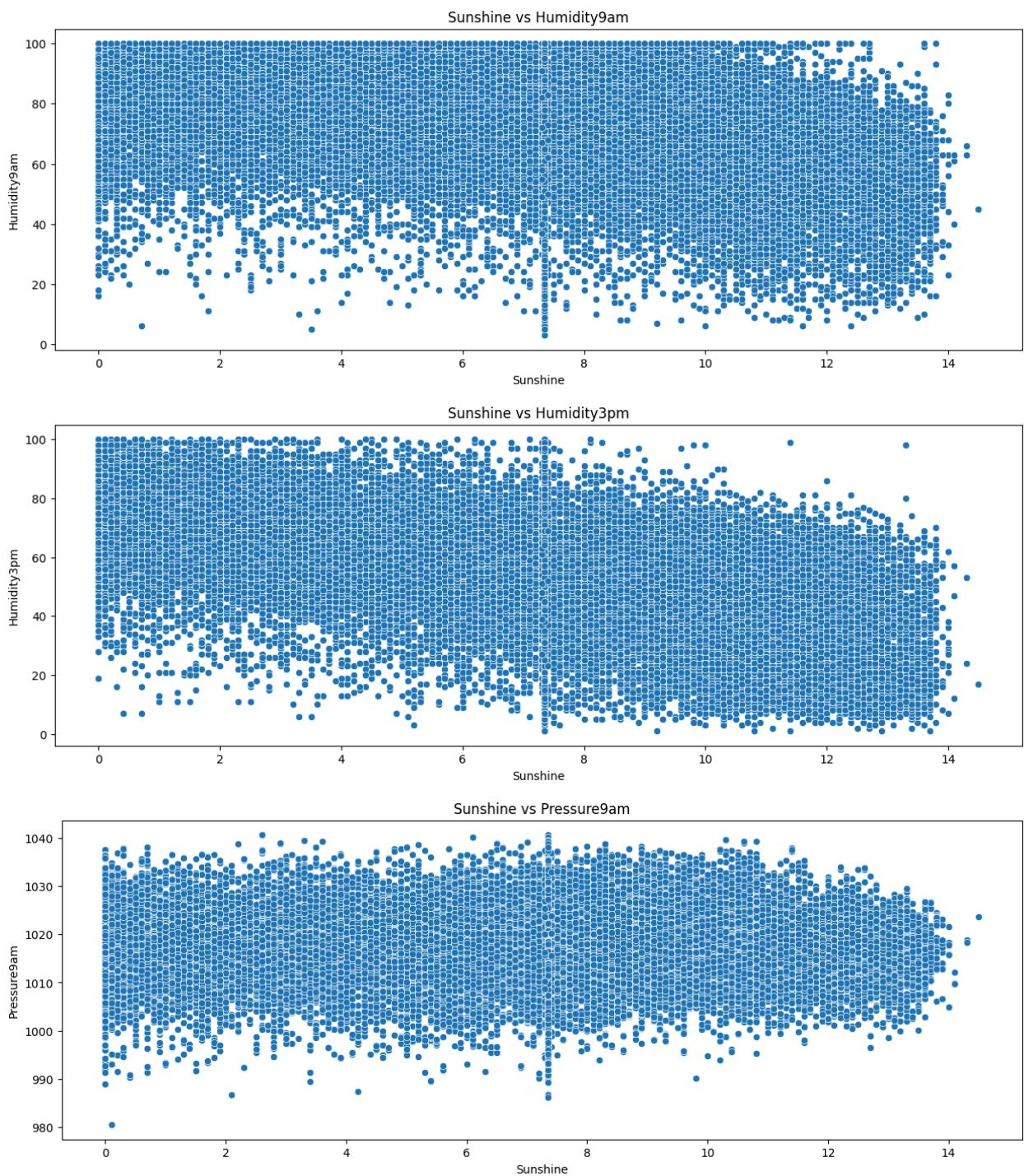
Evaporation vs Pressure9am

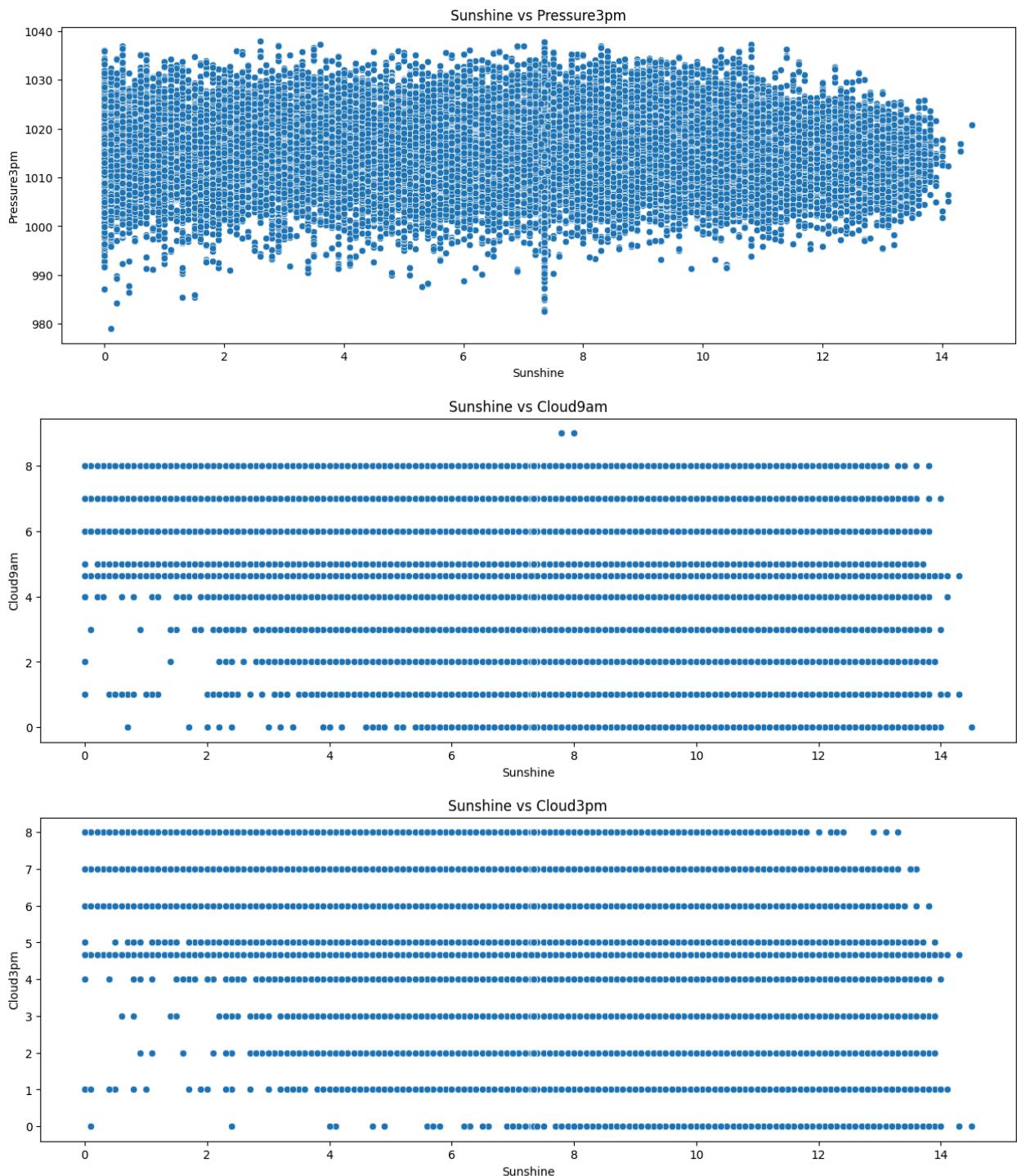


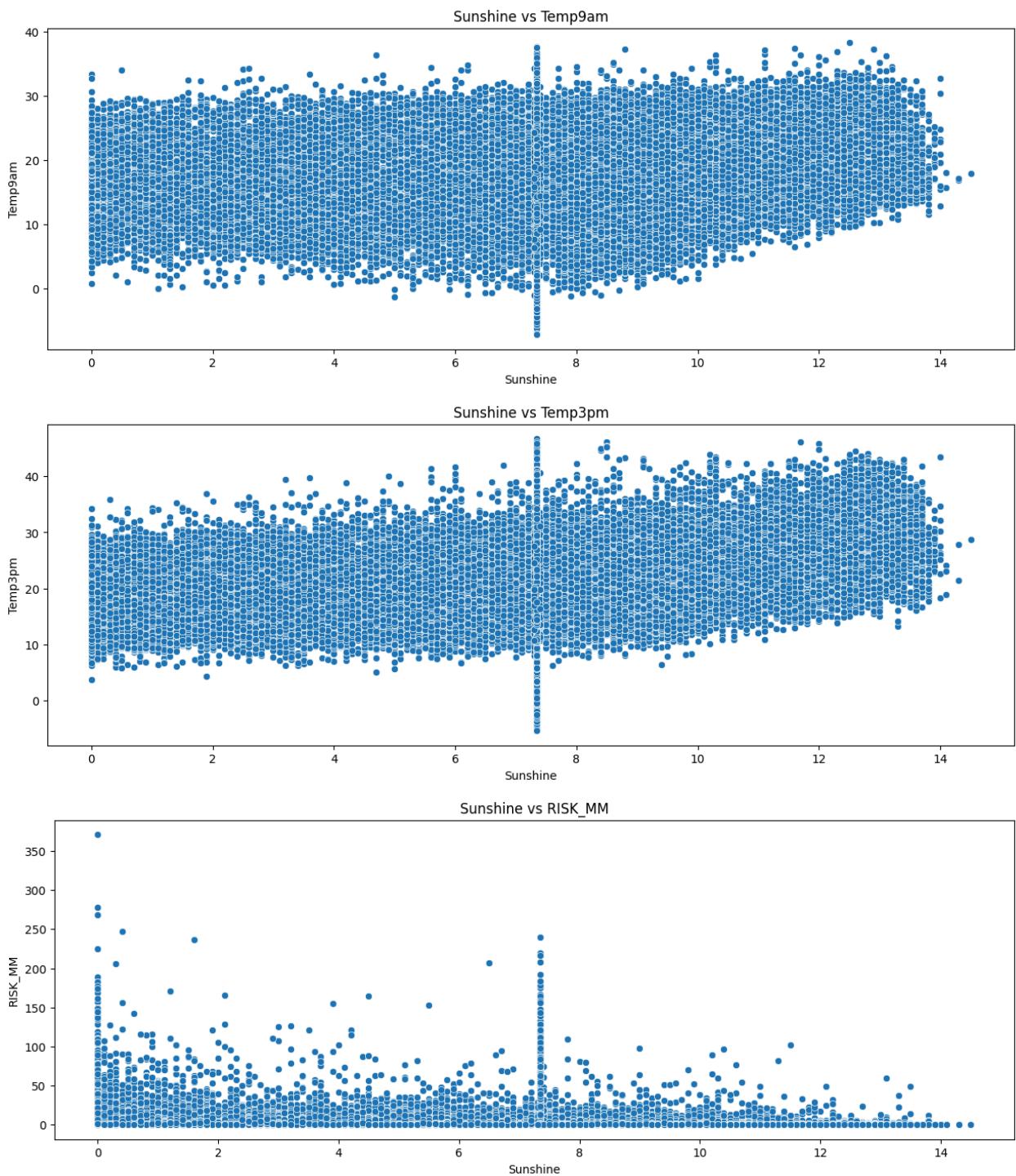




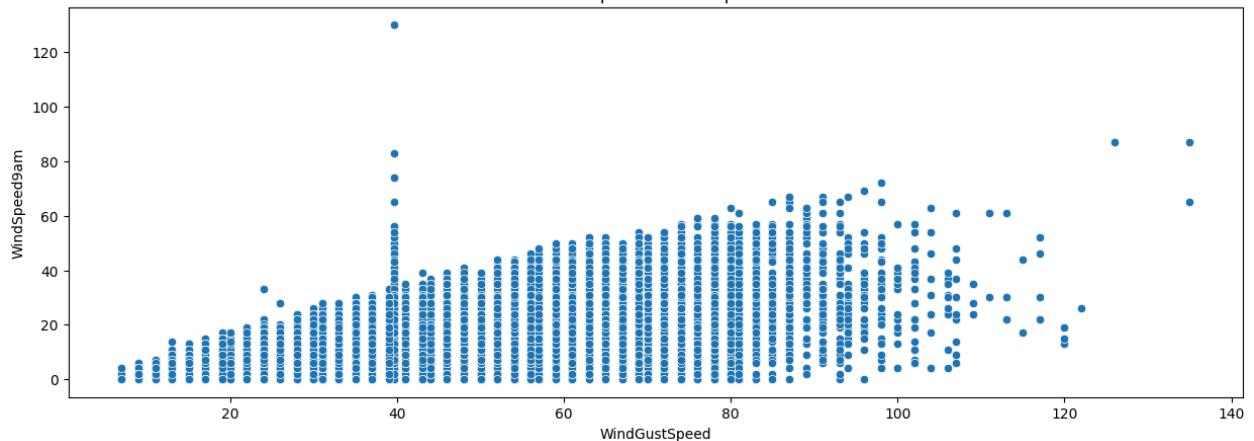




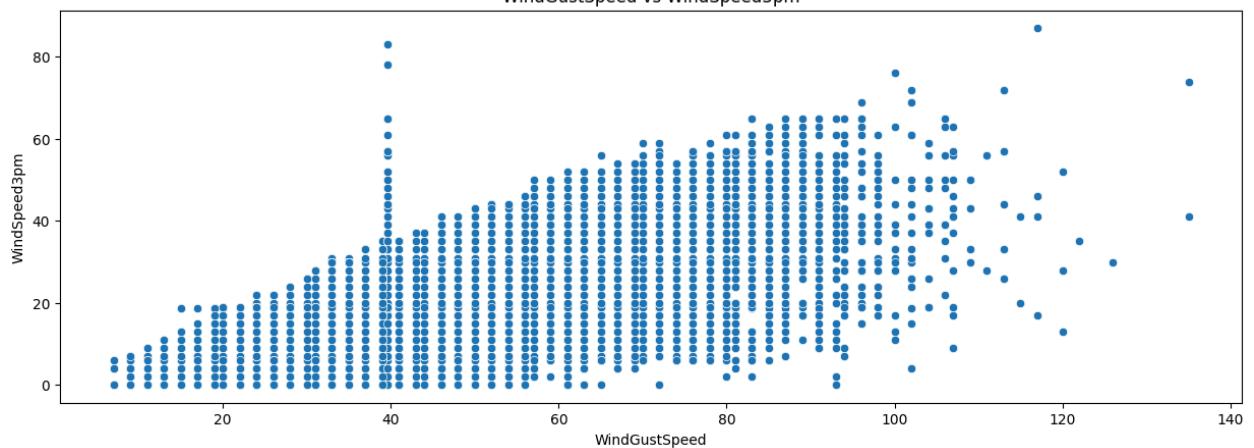




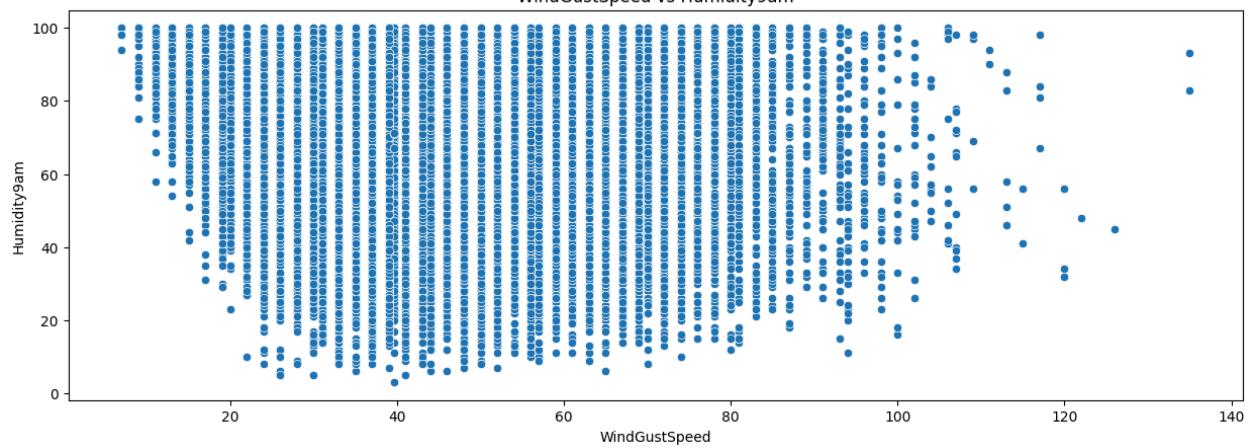
WindGustSpeed vs WindSpeed9am

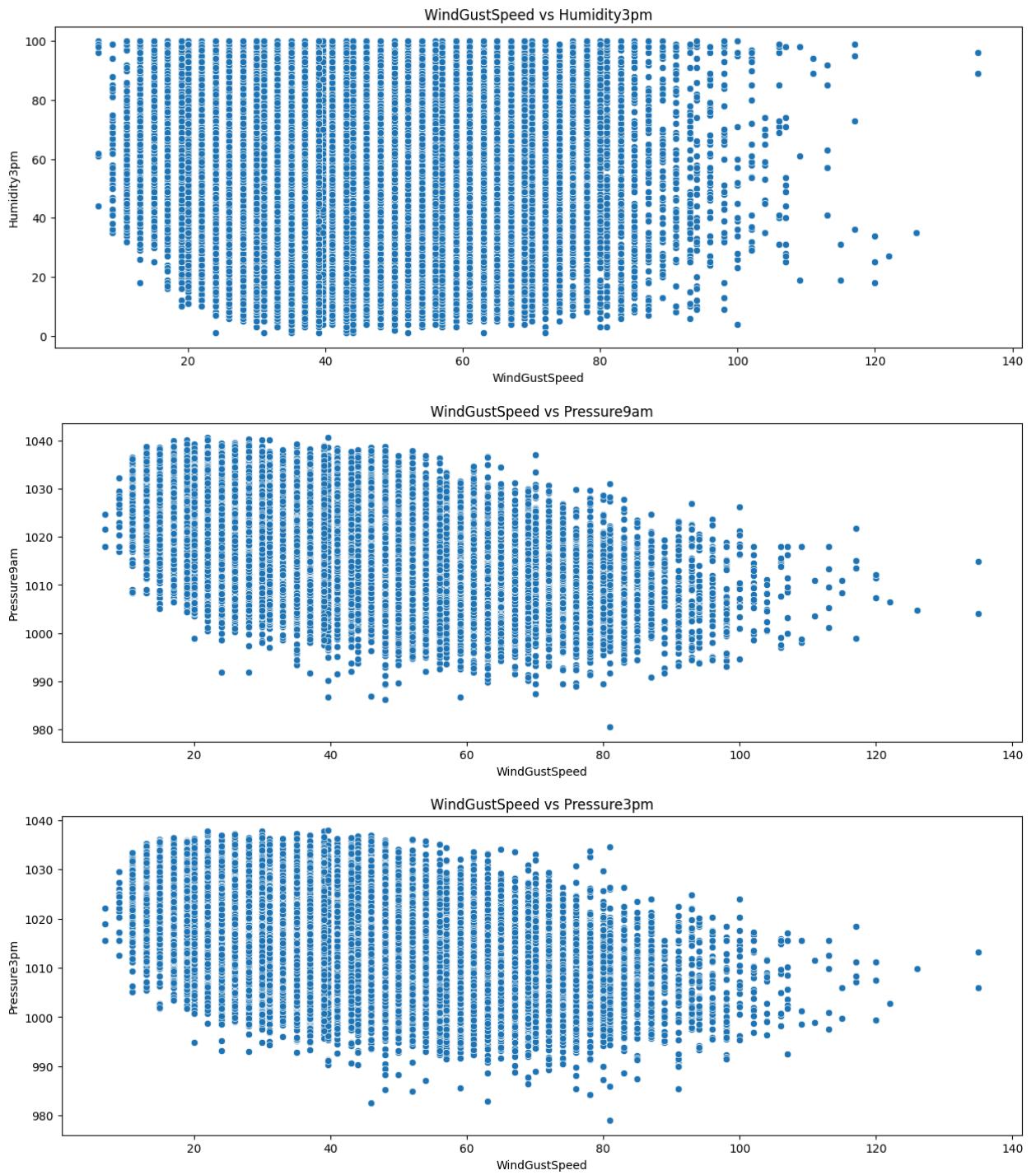


WindGustSpeed vs WindSpeed3pm

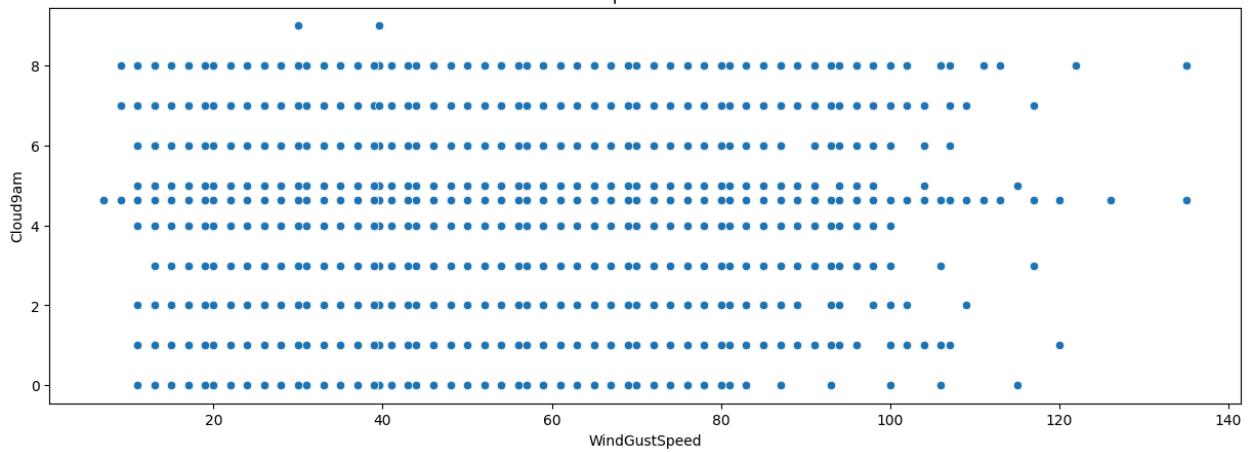


WindGustSpeed vs Humidity9am

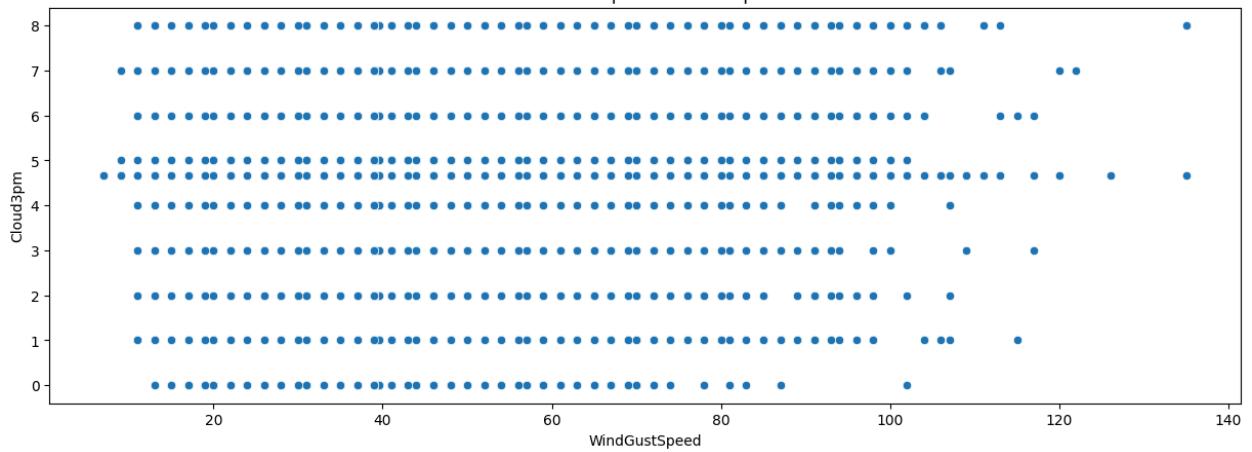




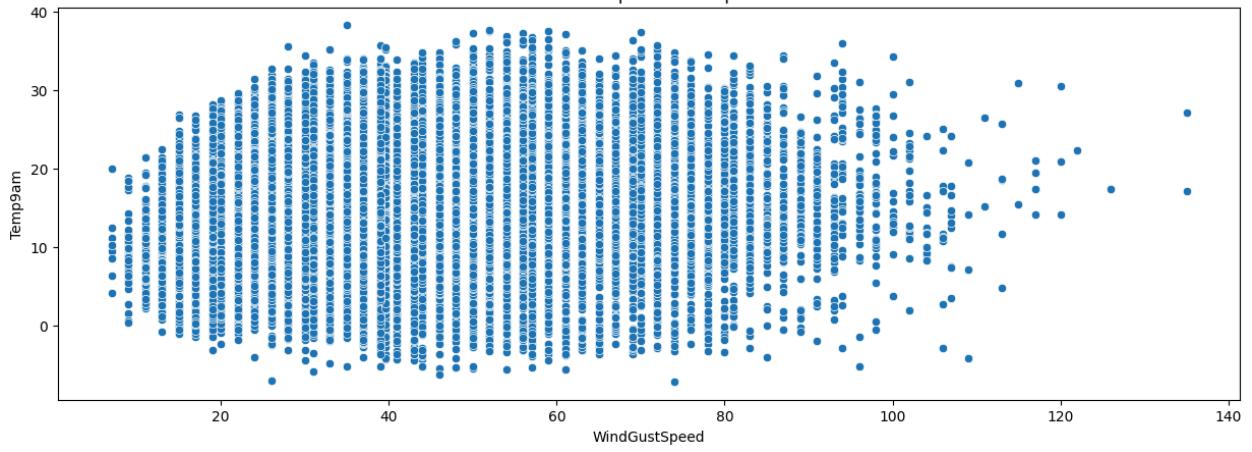
WindGustSpeed vs Cloud9am



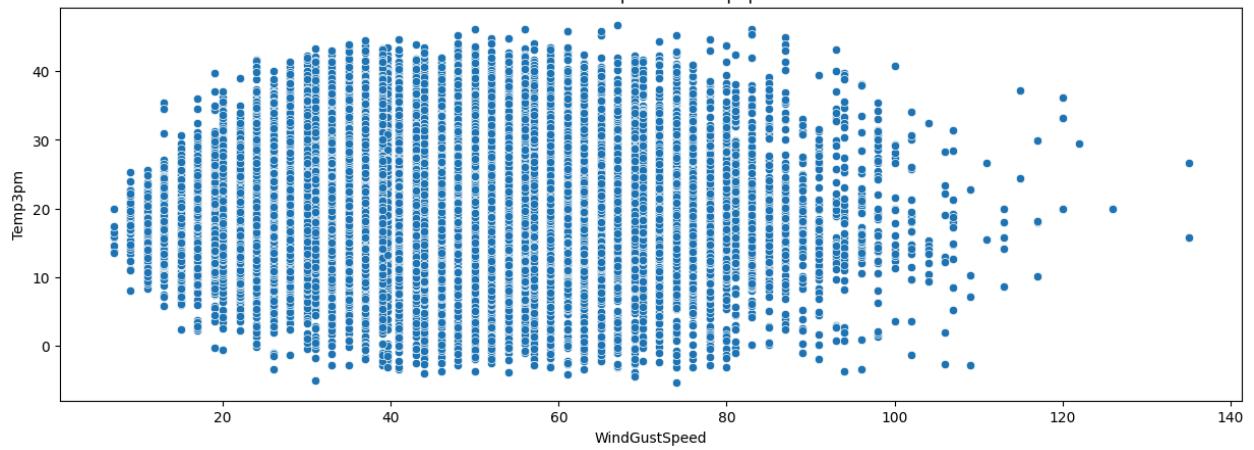
WindGustSpeed vs Cloud3pm



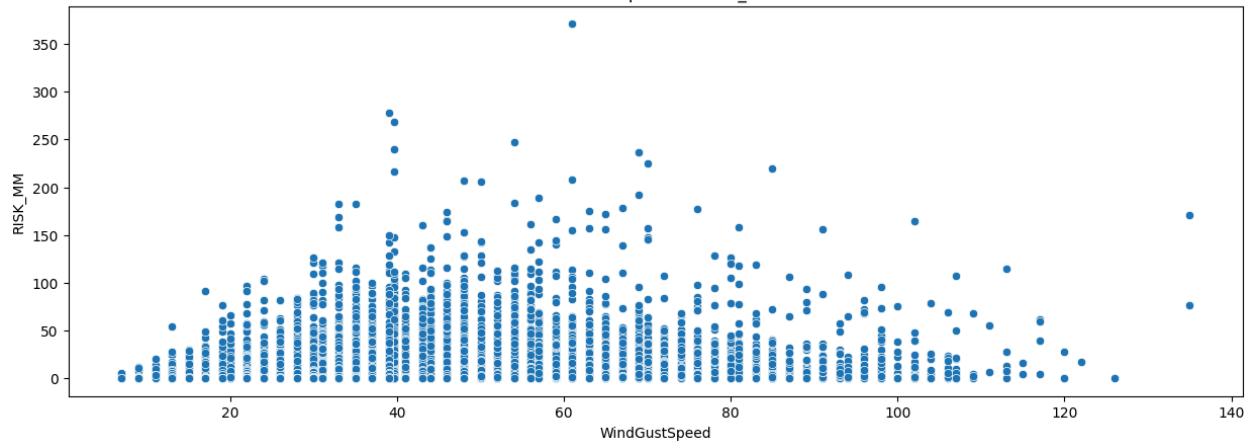
WindGustSpeed vs Temp9am



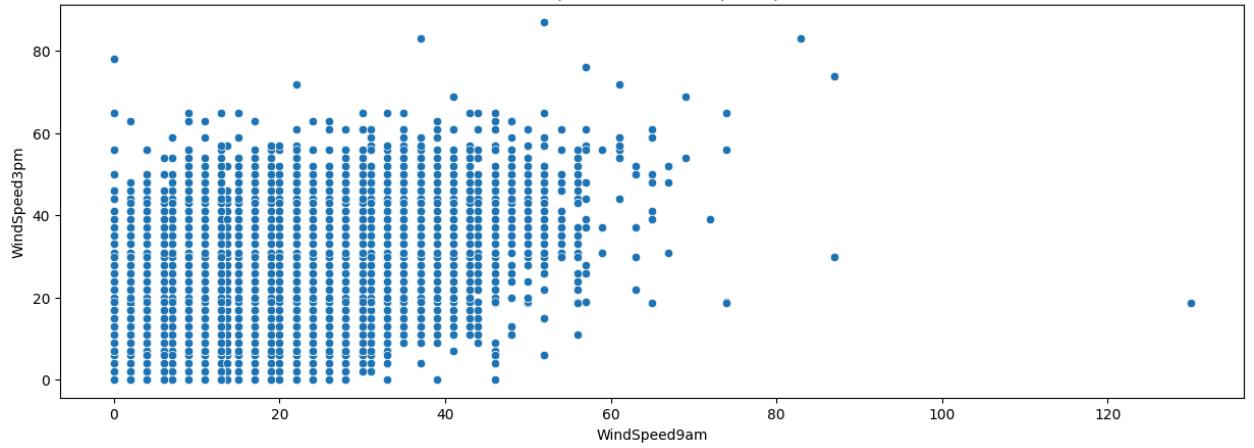
WindGustSpeed vs Temp3pm

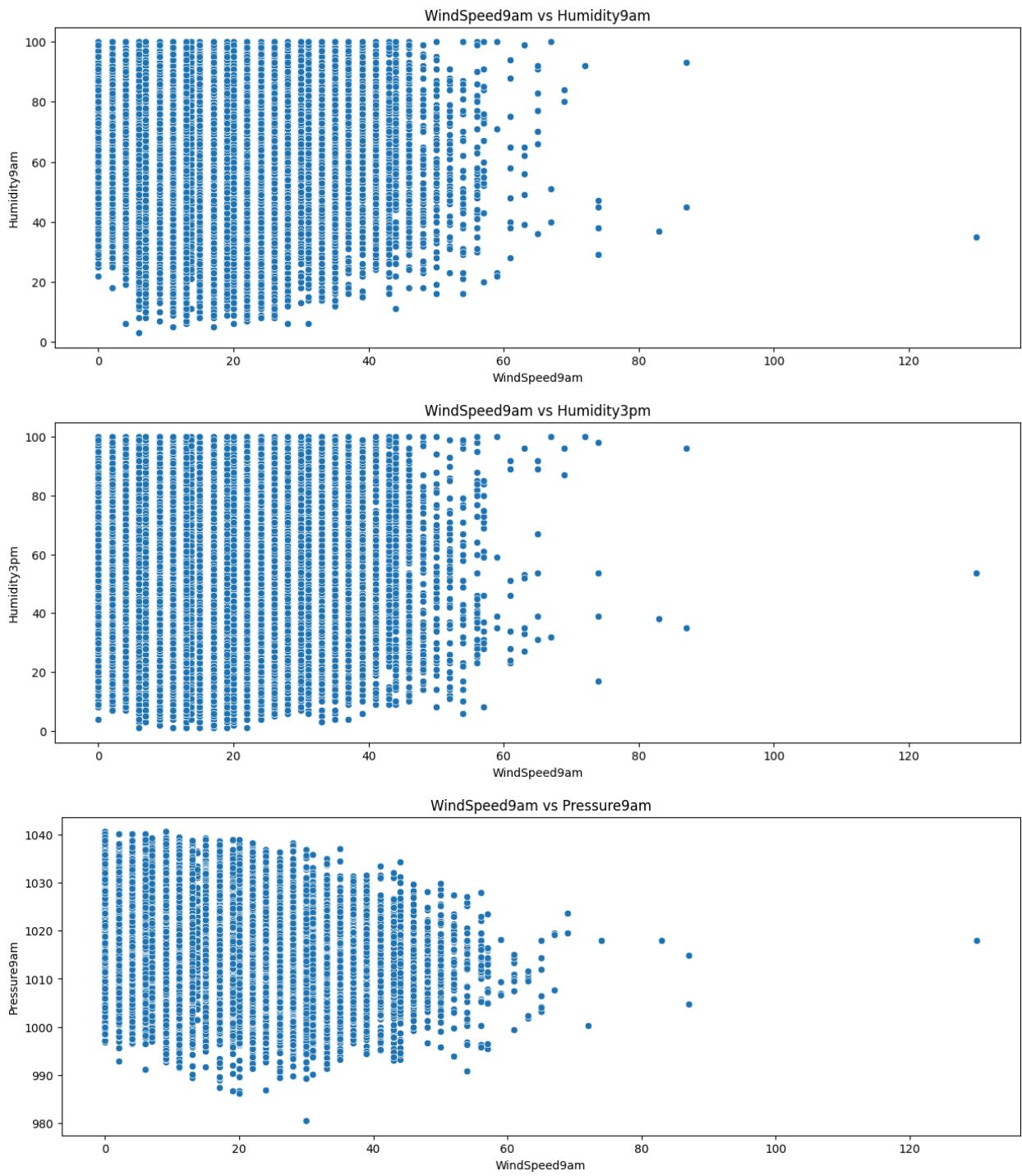


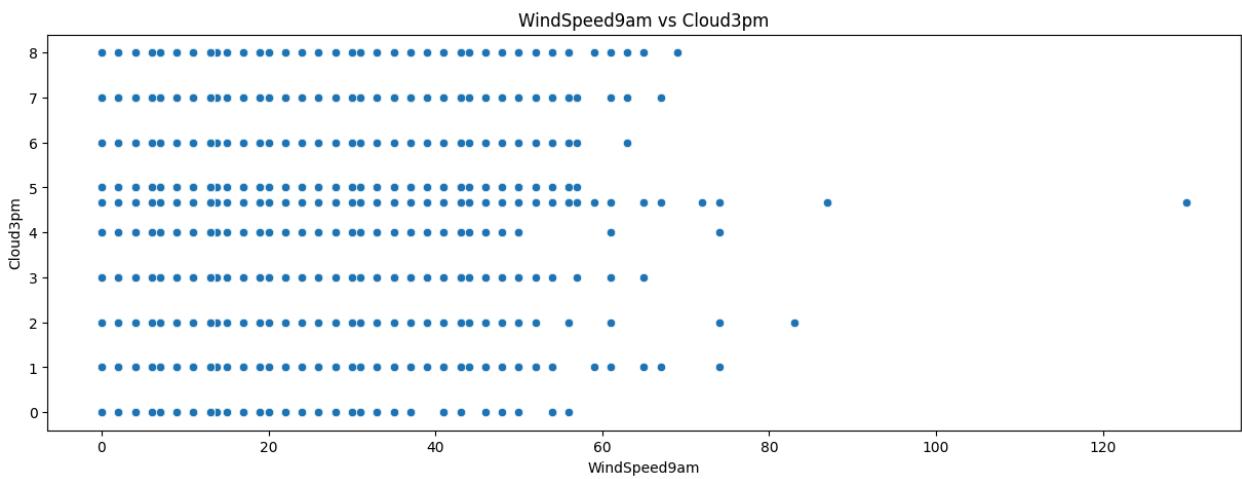
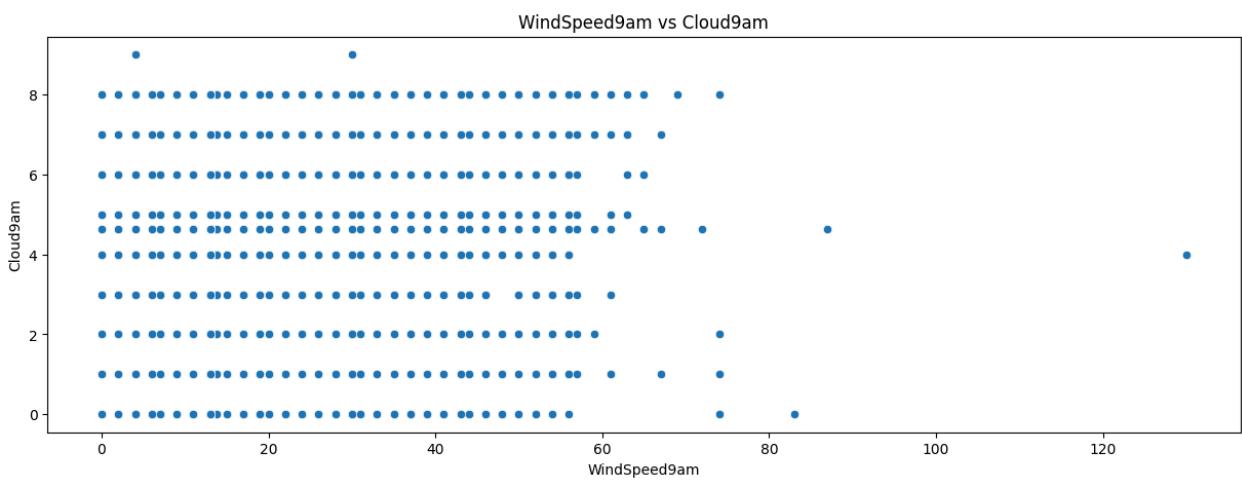
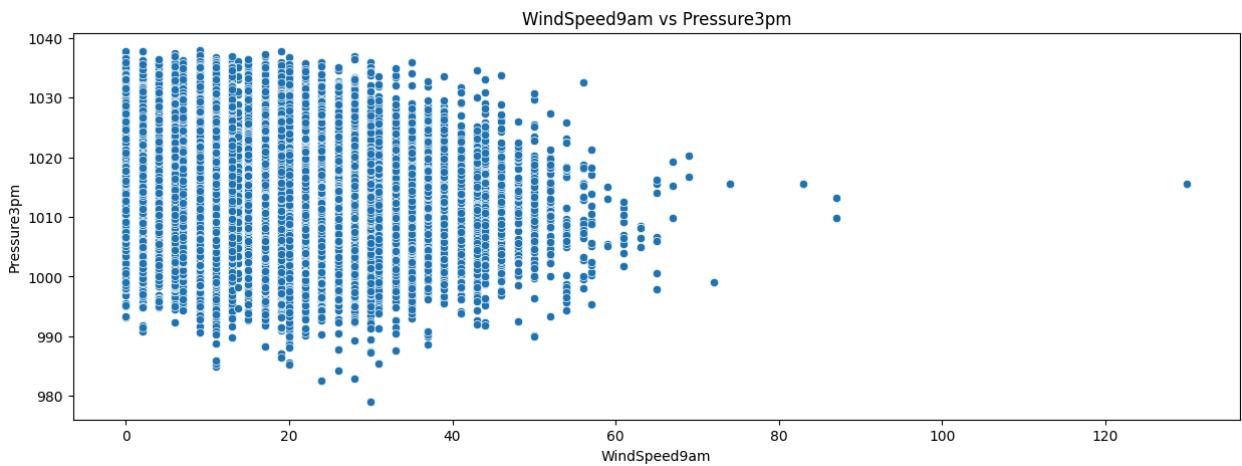
WindGustSpeed vs RISK\_MM

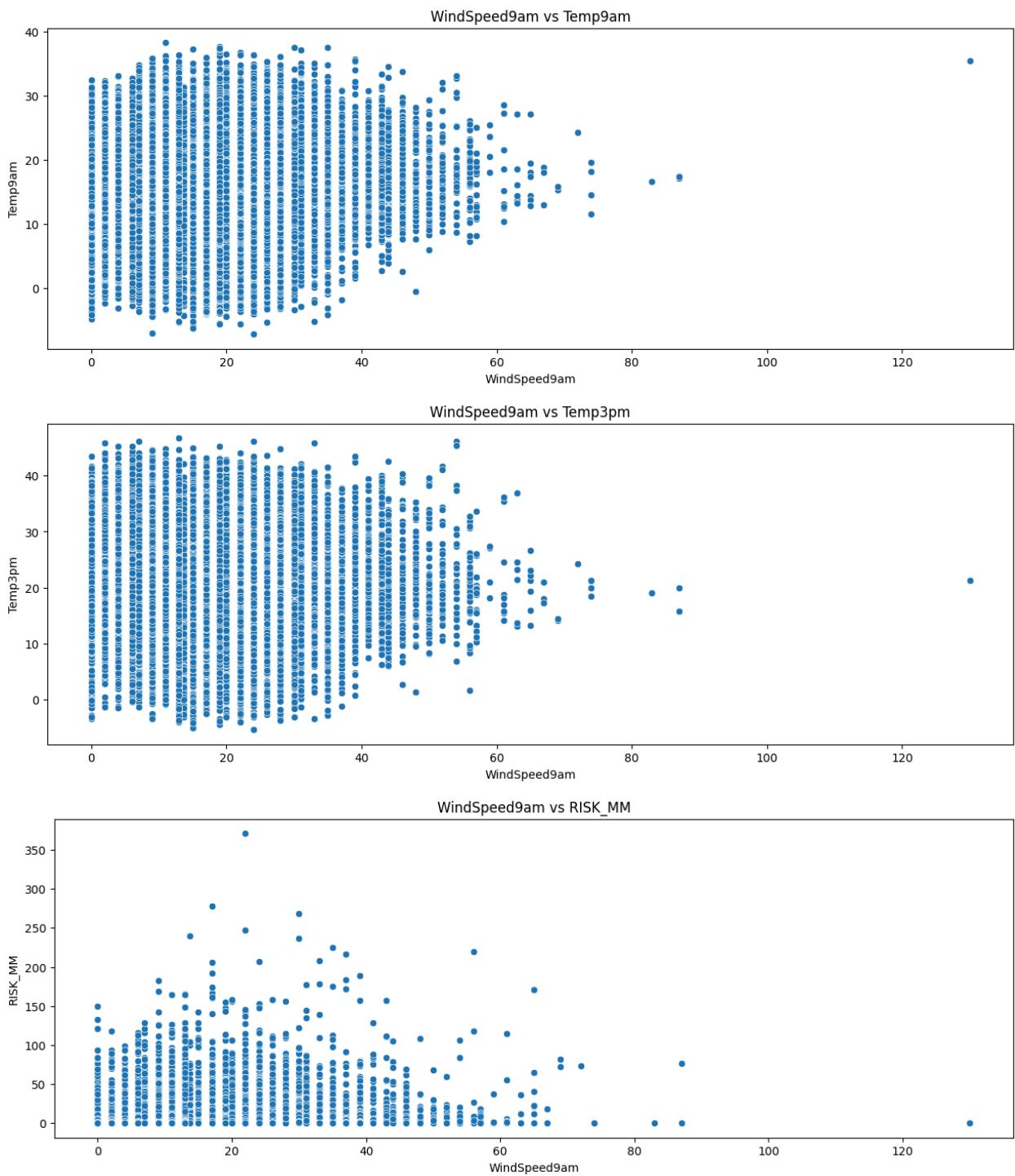


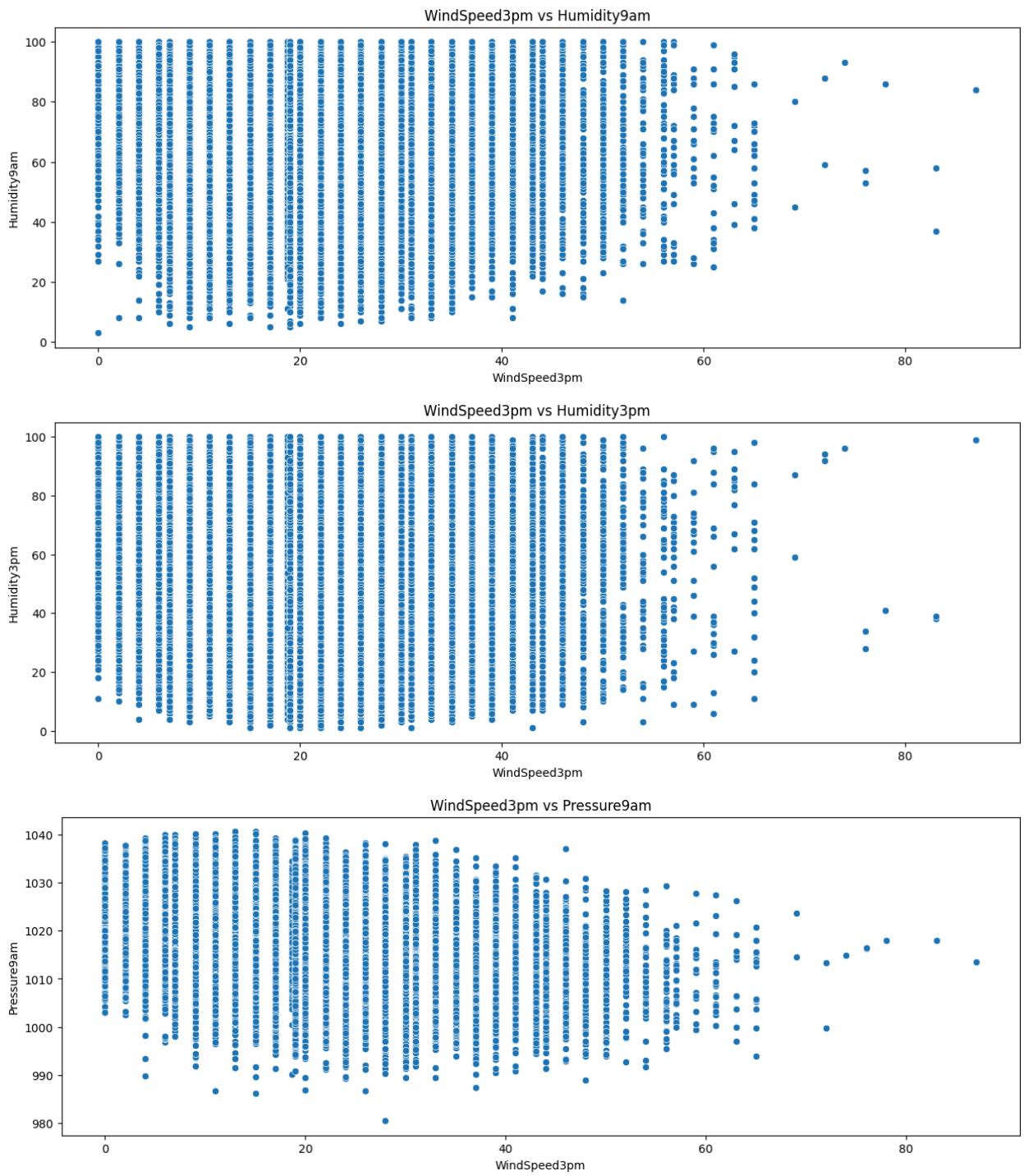
WindSpeed9am vs WindSpeed3pm

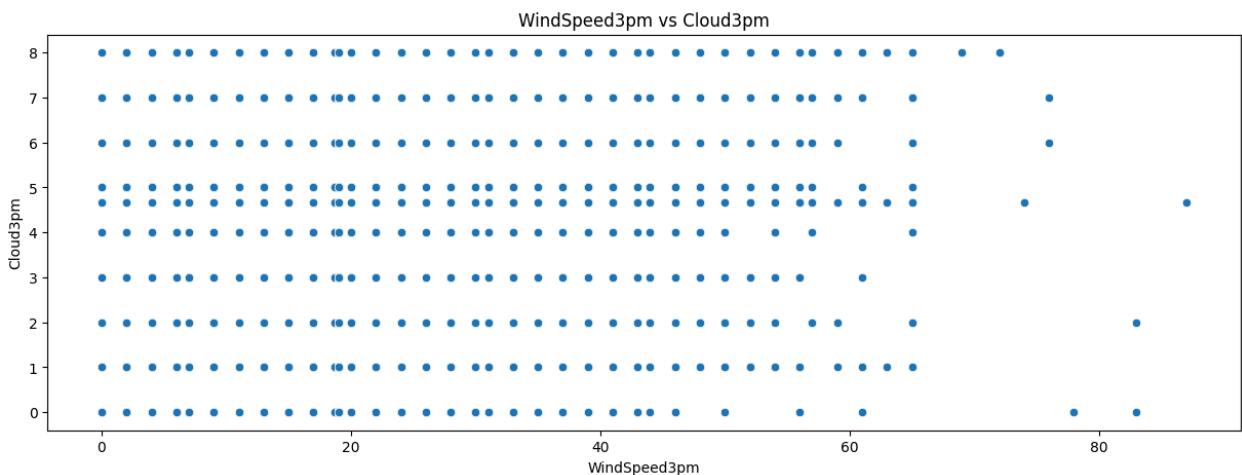
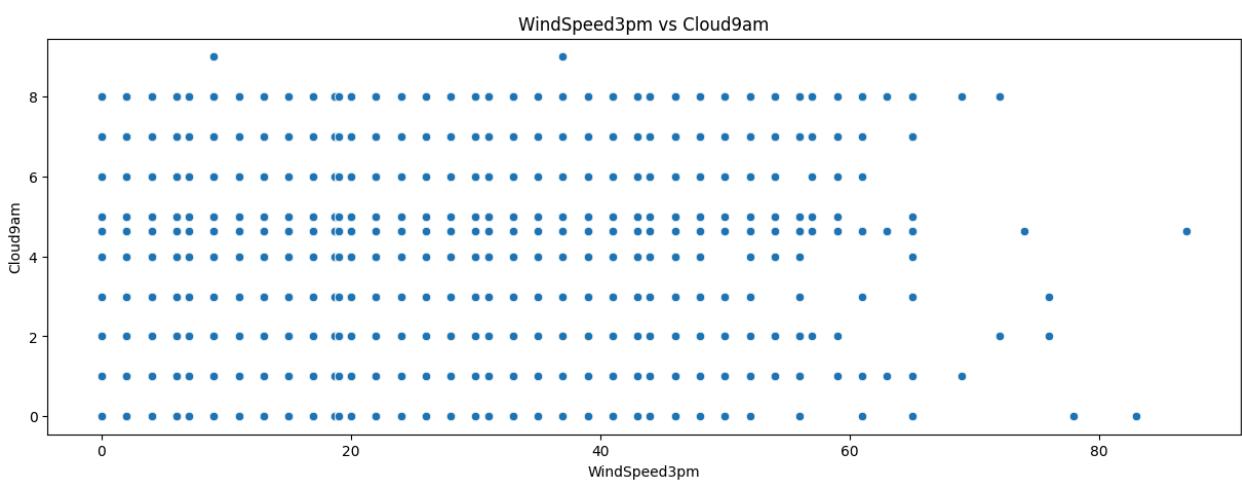
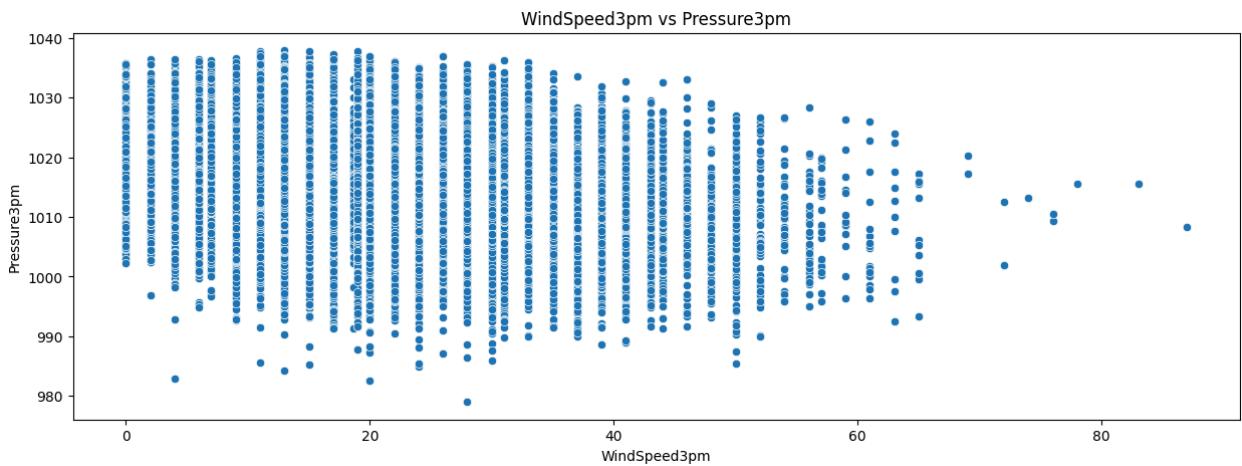


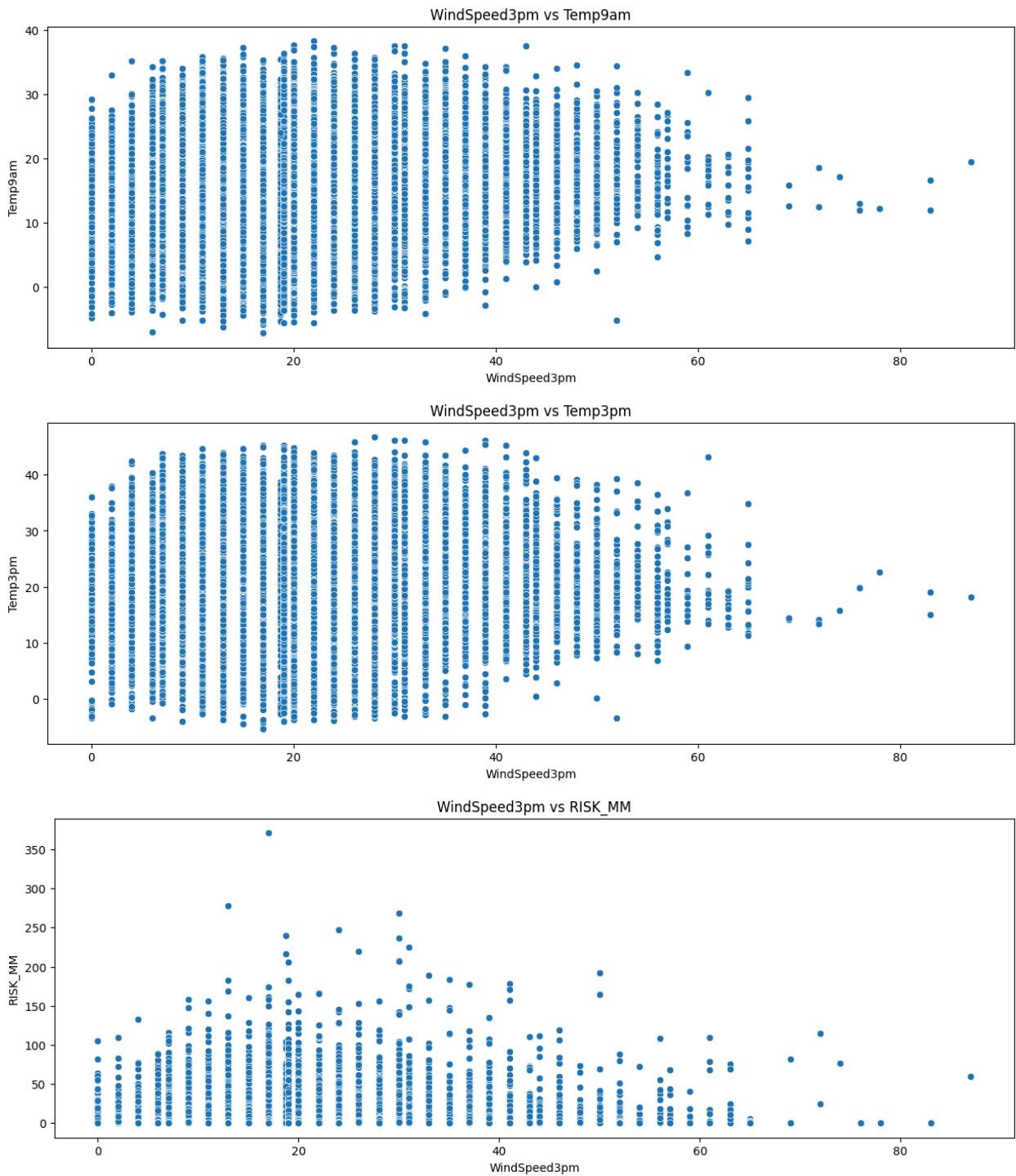


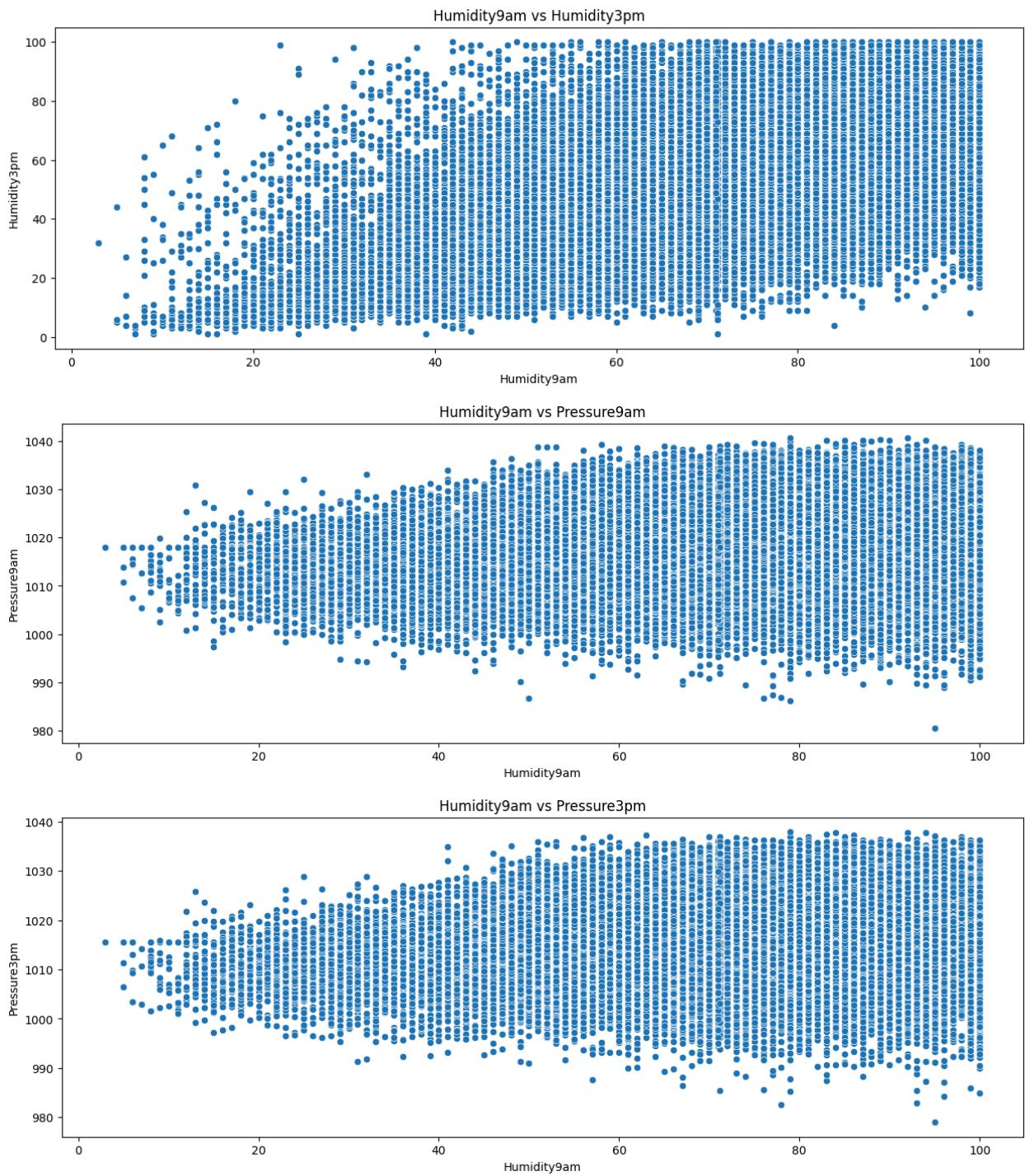




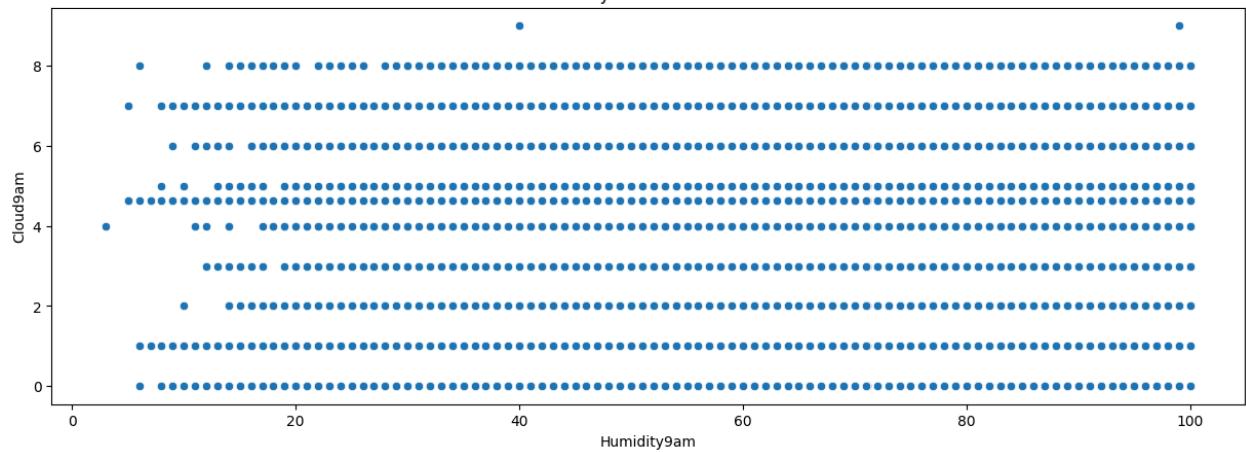




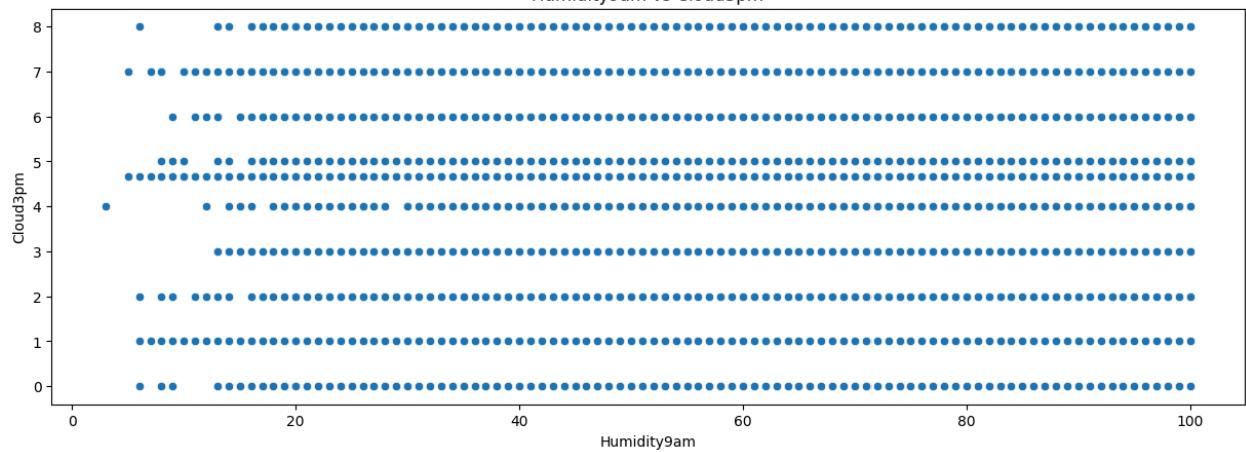




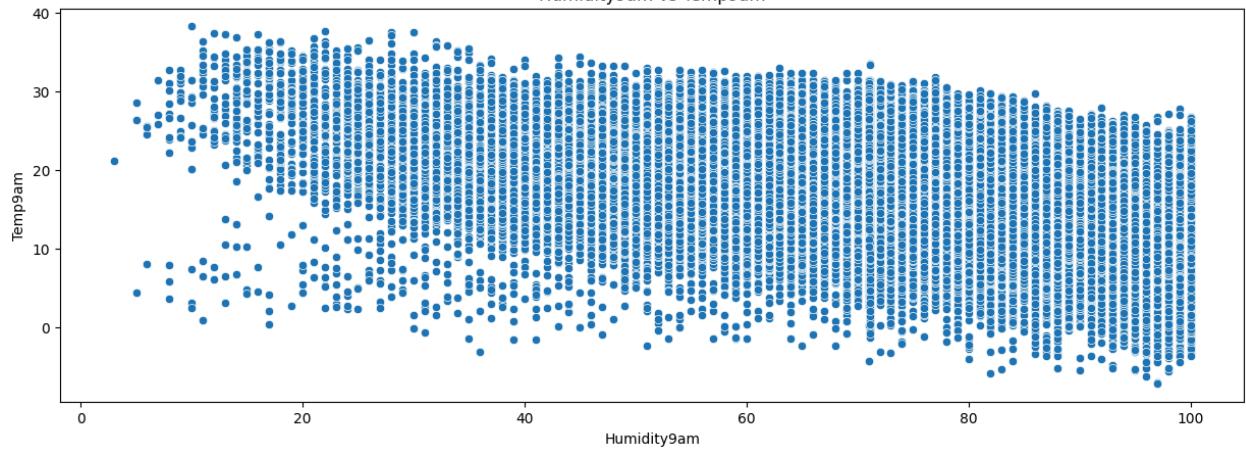
Humidity9am vs Cloud9am



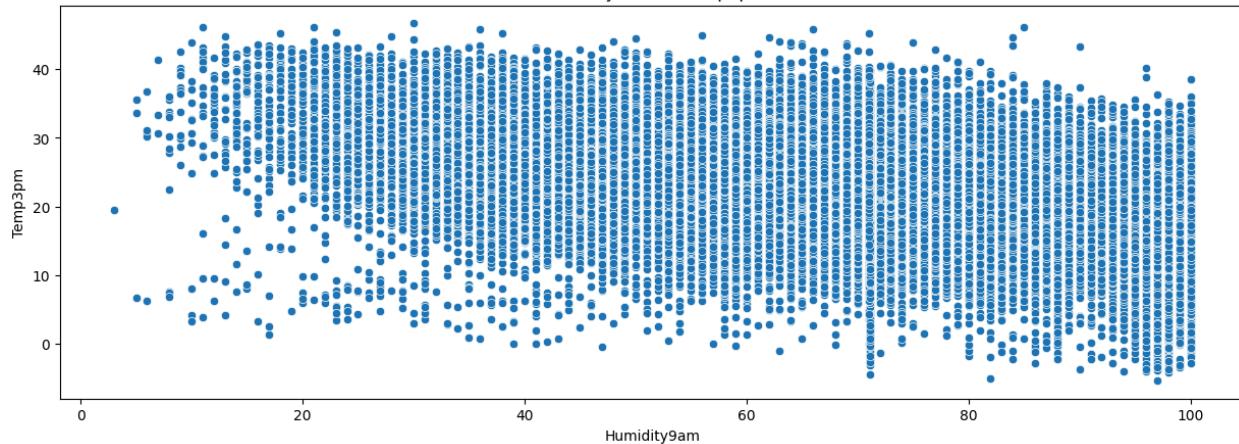
Humidity9am vs Cloud3pm



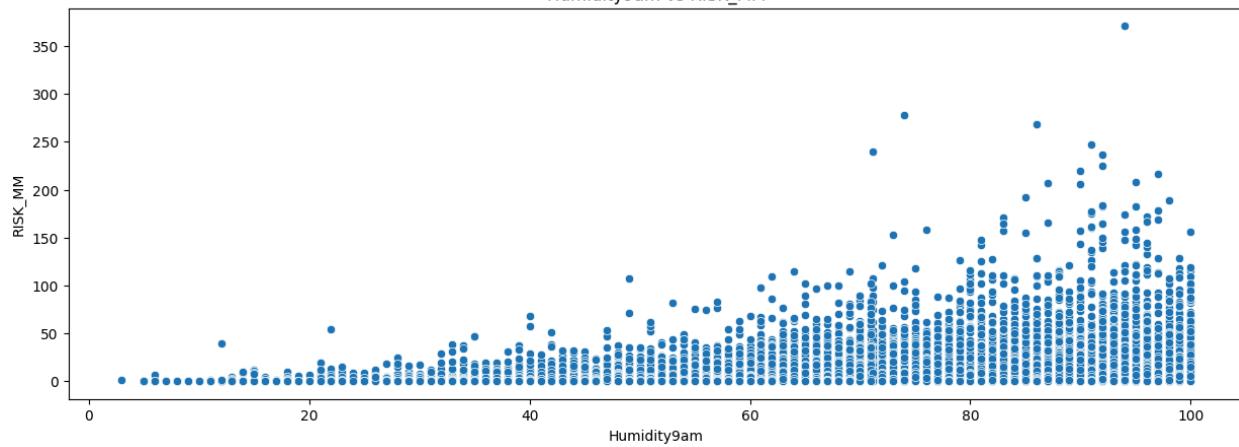
Humidity9am vs Temp9am



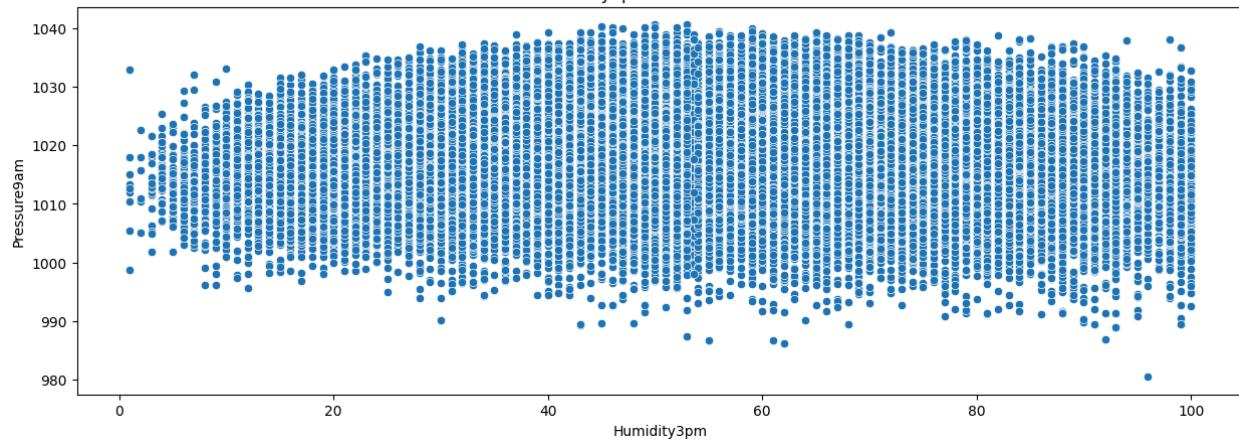
Humidity9am vs Temp3pm

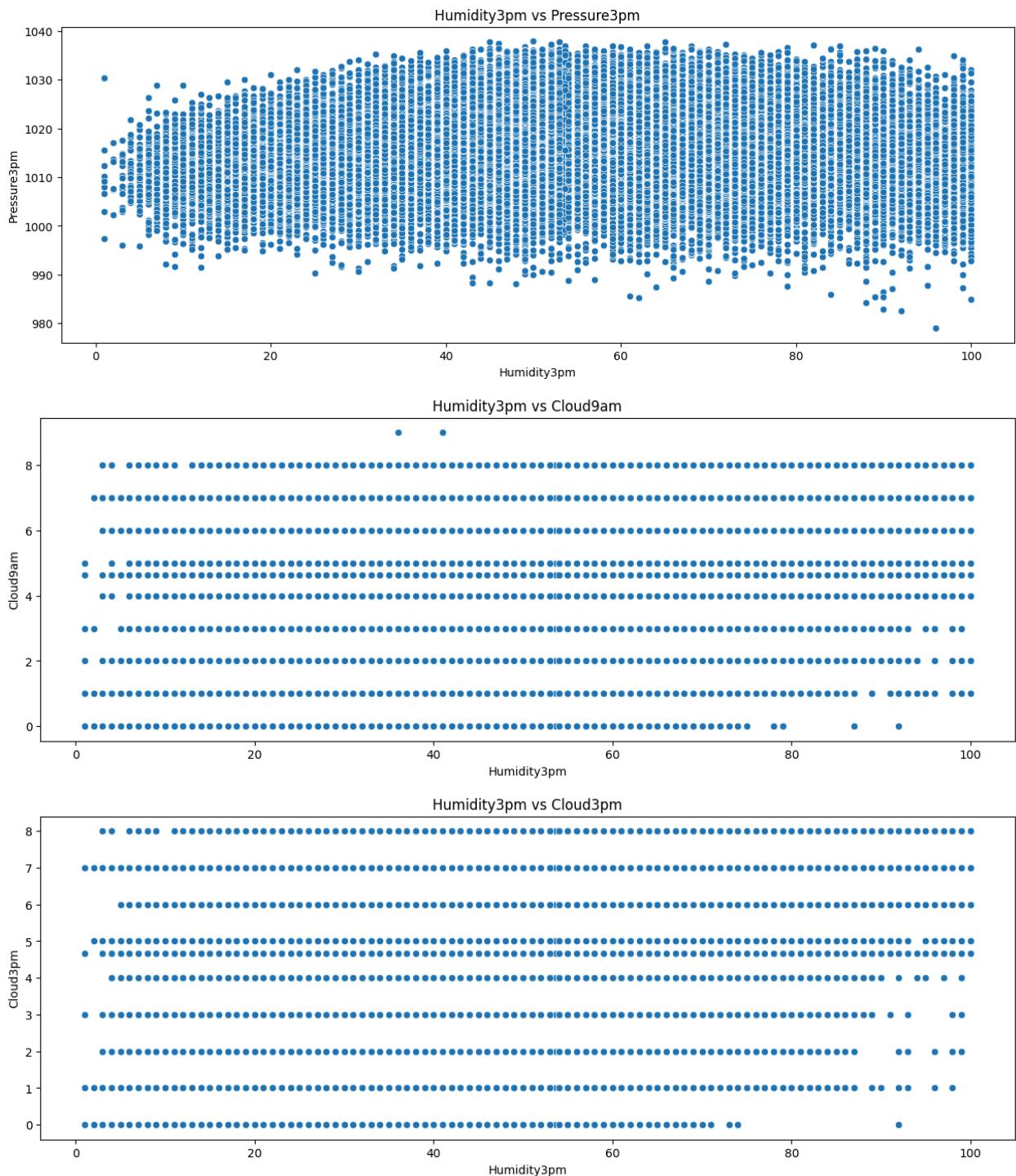


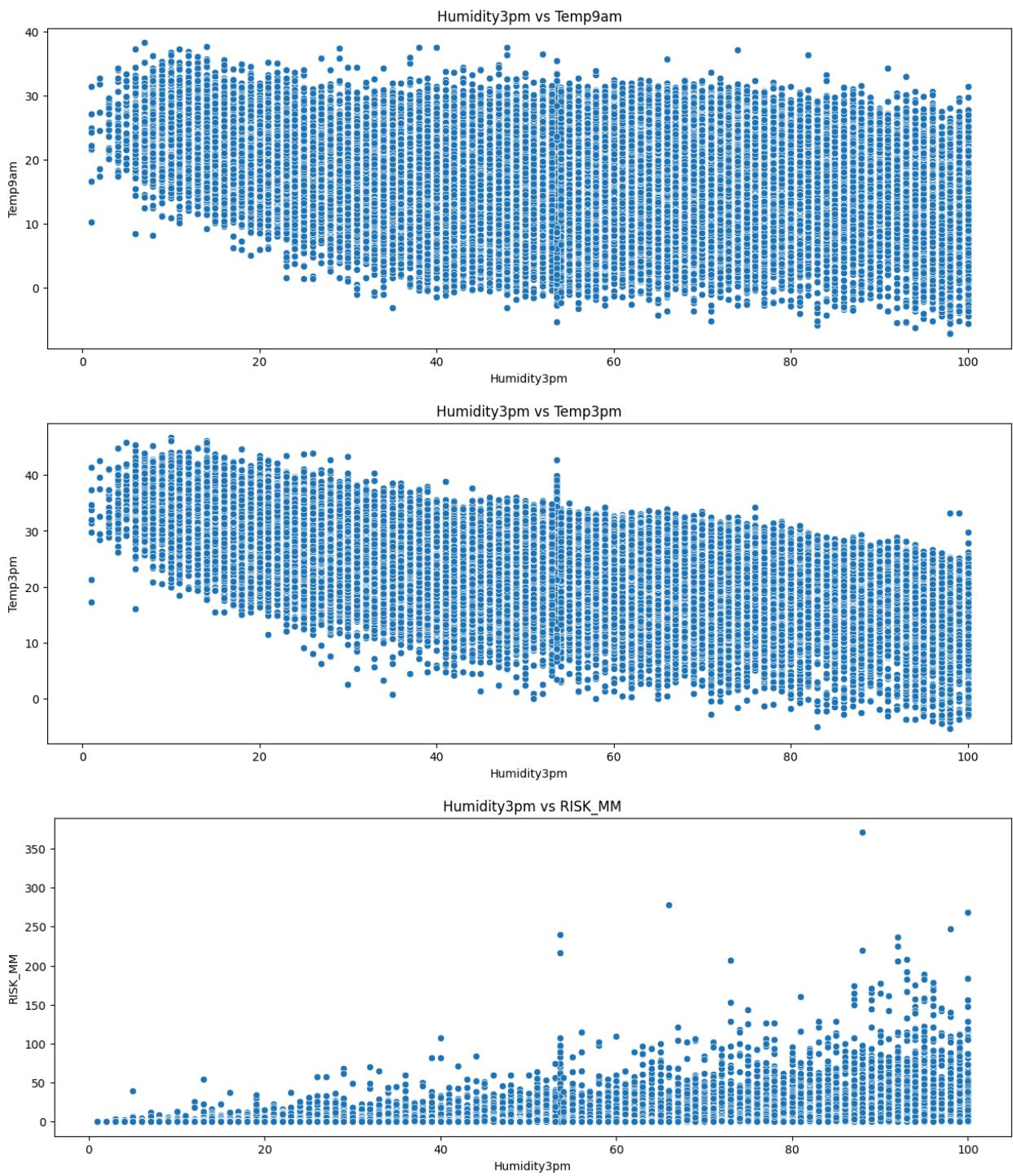
Humidity9am vs RISK\_MM

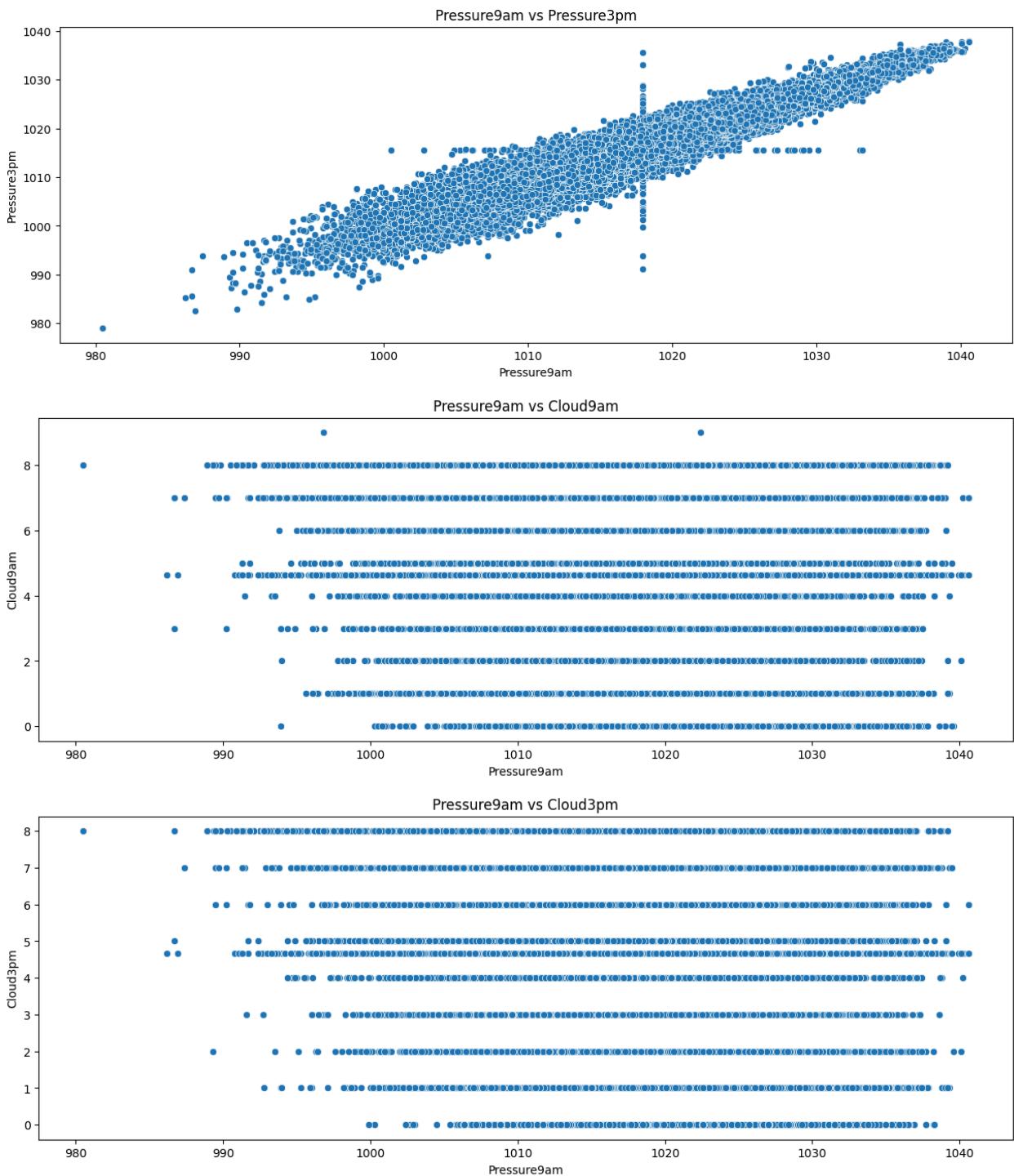


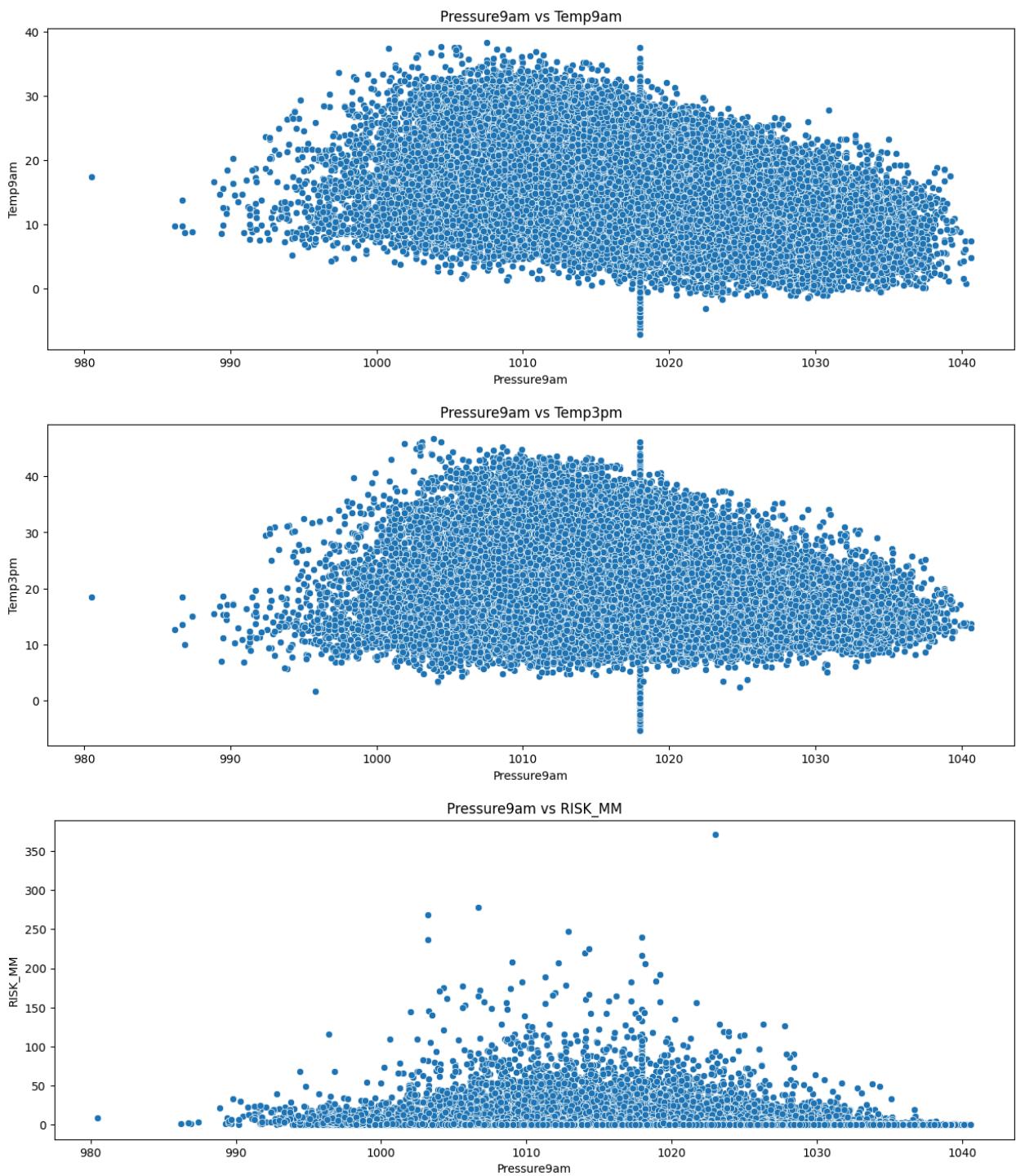
Humidity3pm vs Pressure9am

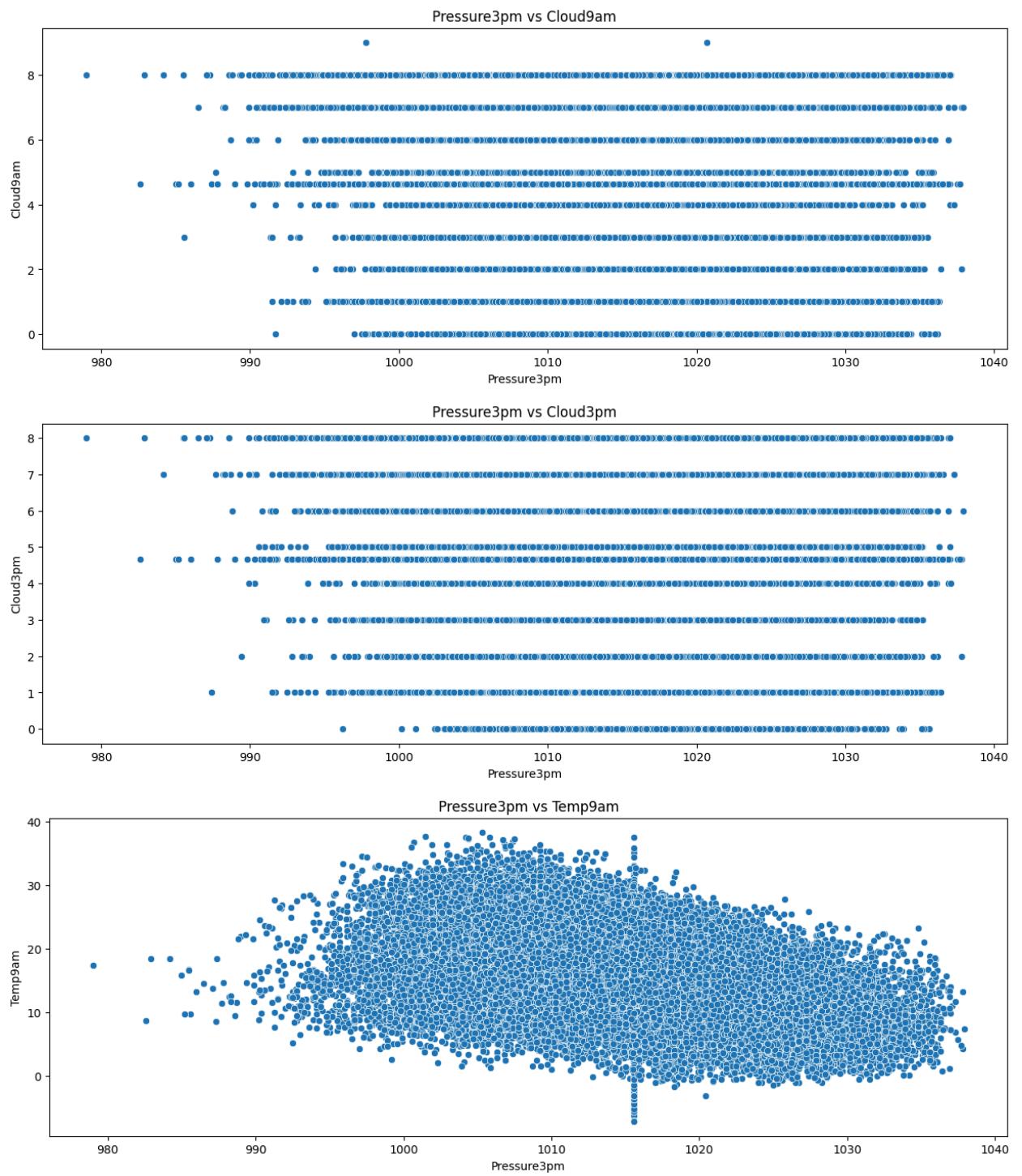


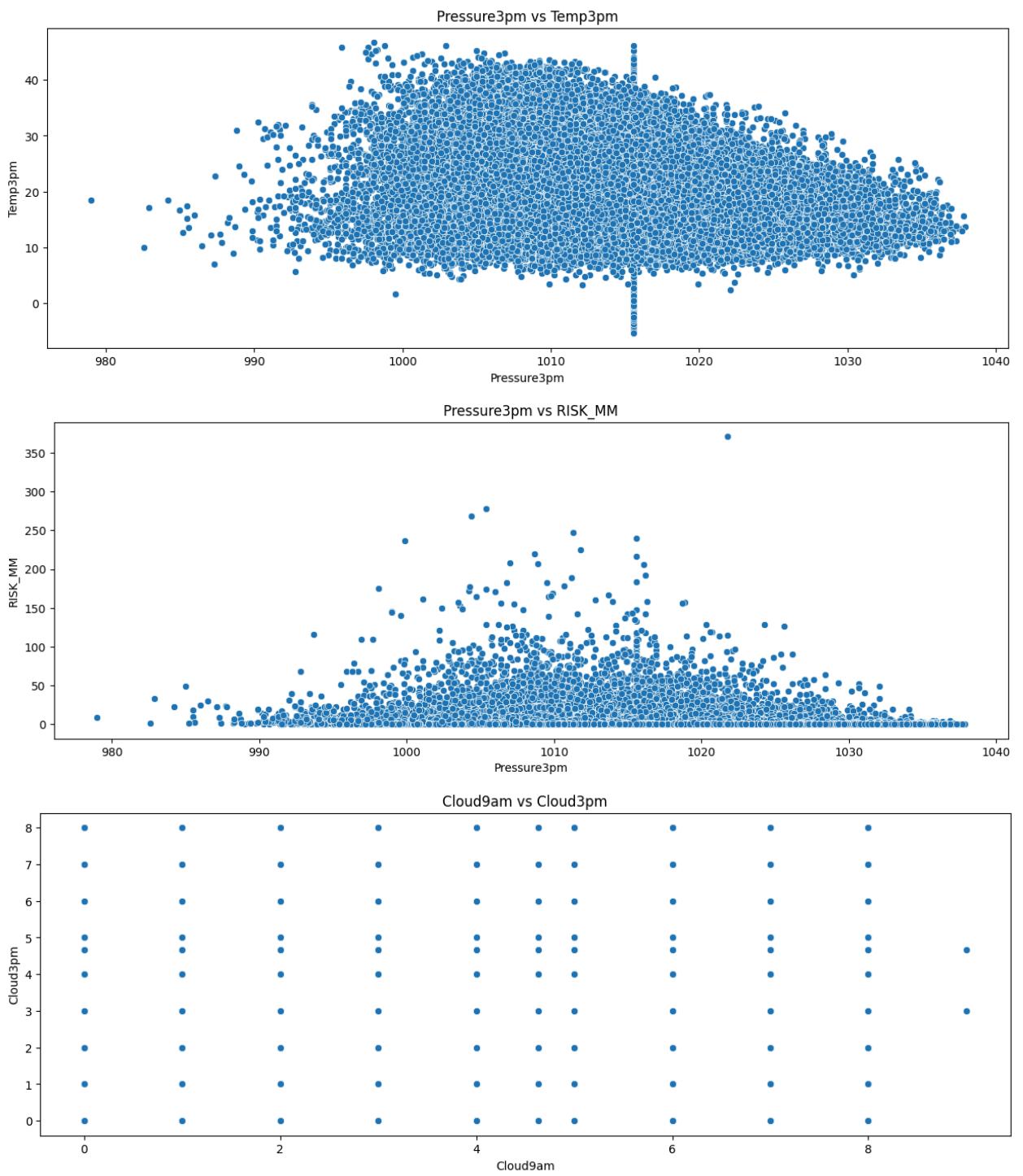


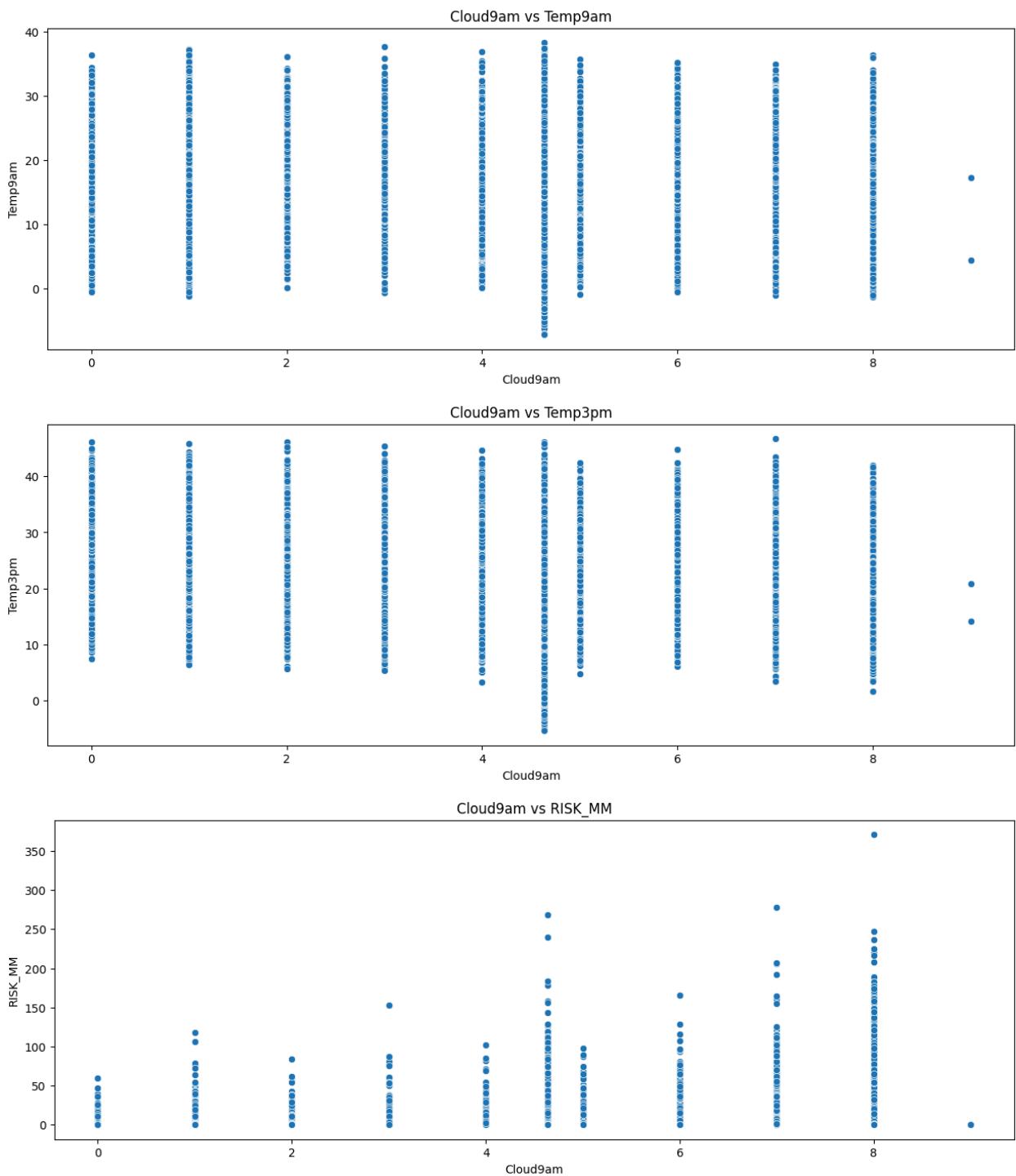


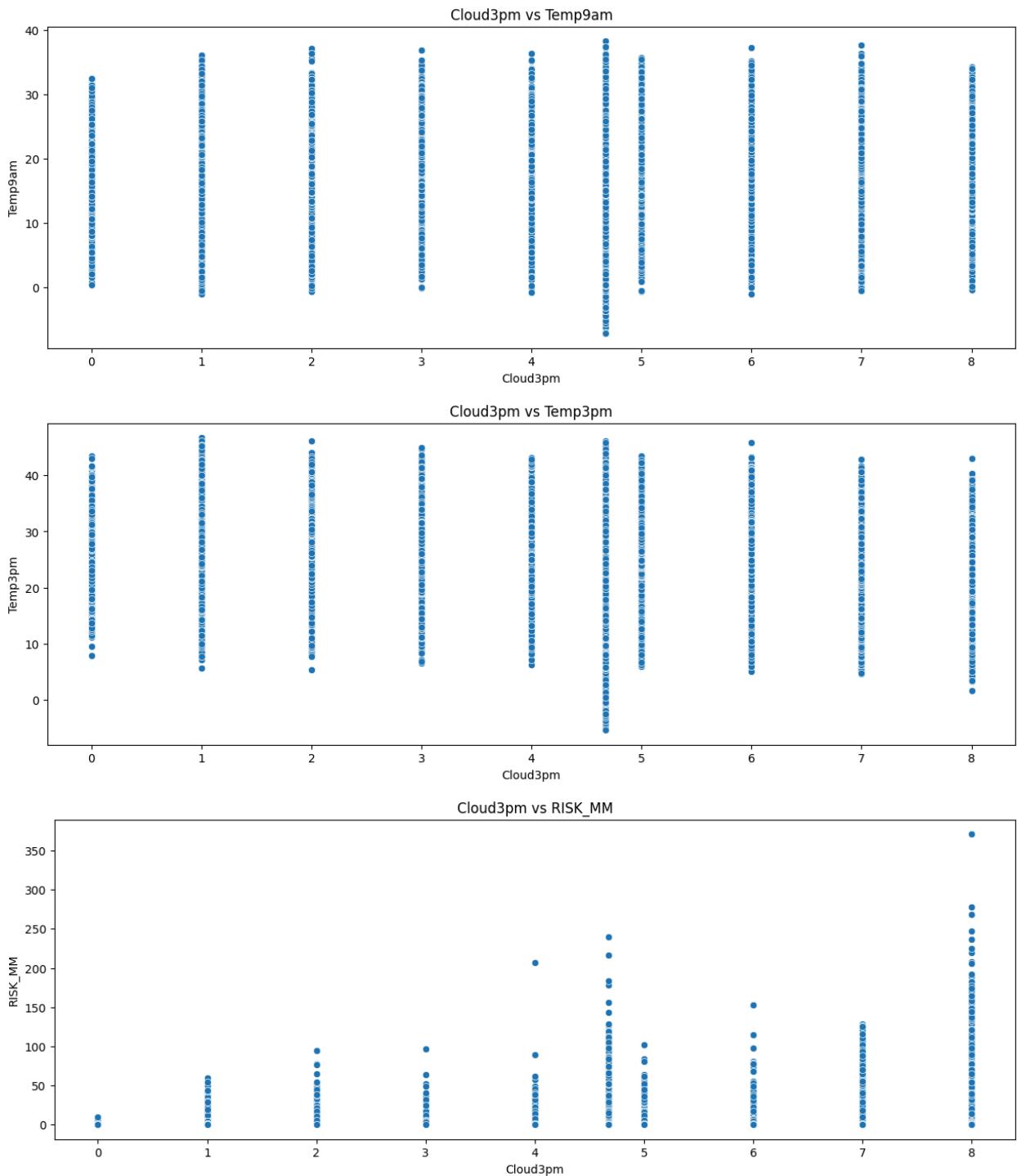


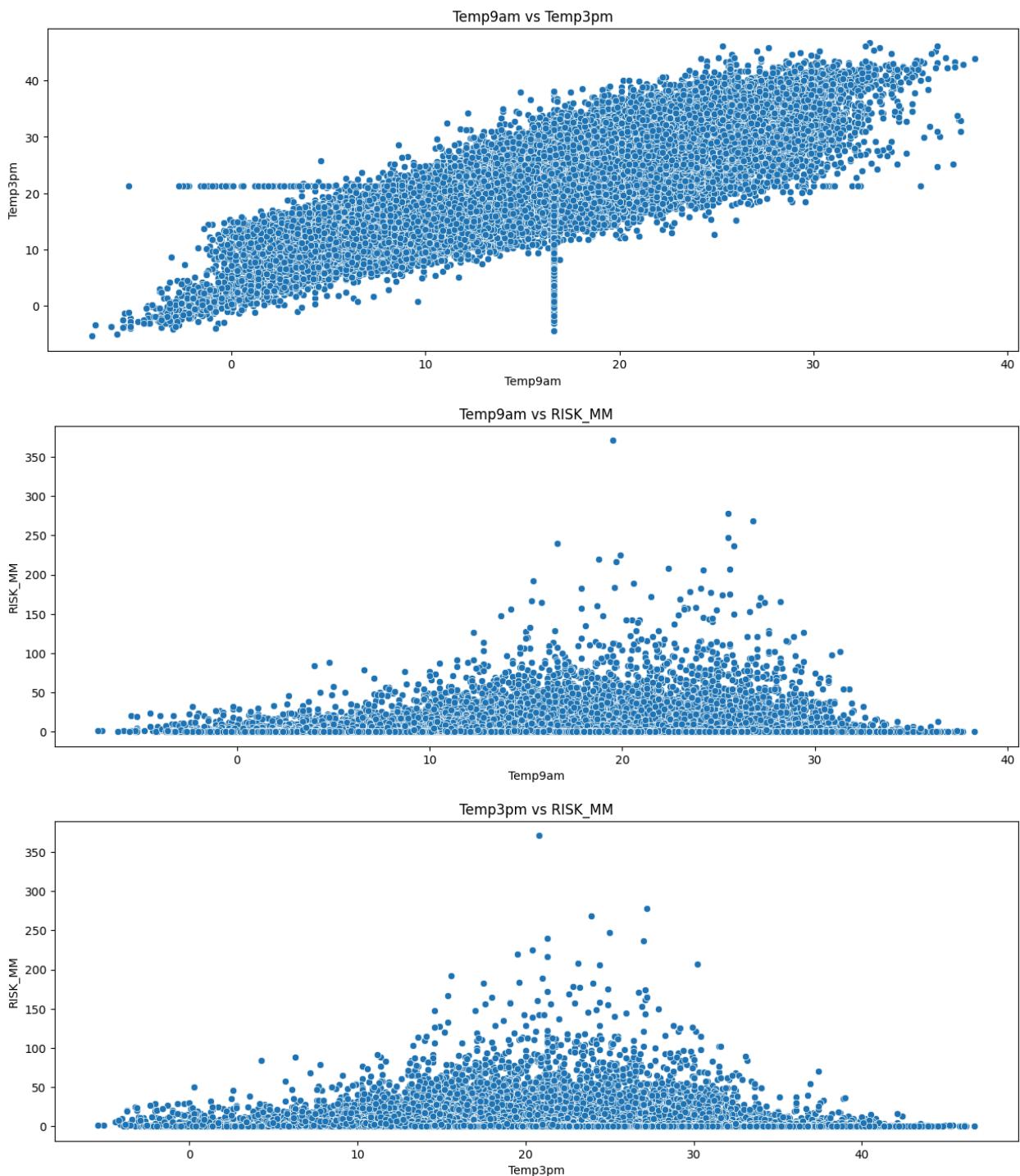












```
In [ ]: def remove_outliers(df):
    outeres12=set()
    for i in df.select_dtypes(exclude='object'):
        Q1=df[i].quantile(0.25)
        Q3=df[i].quantile(0.75)
        IQR=Q3-Q1
        lower=Q1-1.5*IQR
        upper=Q3+1.5*IQR
        outliers=df[(df[i]<lower) | (df[i]>upper)].index
        outeres12.update(outliers)
```

```
    return df.drop(index=outeres12)
df5_clean=remove_outliers(df5)

In [ ]: df5_clean['Date']=pd.to_datetime(df5_clean['Date'])

In [ ]: df5_clean['RainTomorrow']=df5_clean['RainTomorrow'].str.replace('No','0')
df5_clean['RainTomorrow']=df5_clean['RainTomorrow'].str.replace('Yes','1')

In [ ]: df5_clean['RainToday']=df5_clean['RainToday'].str.replace('No','0')
df5_clean['RainToday']=df5_clean['RainToday'].str.replace('Yes','1')

In [ ]: df5_clean['RainToday']=df5_clean['RainToday'].astype(int)
df5_clean['RainTomorrow']=df5_clean['RainTomorrow'].astype(int)

In [ ]: from sklearn.preprocessing import LabelEncoder

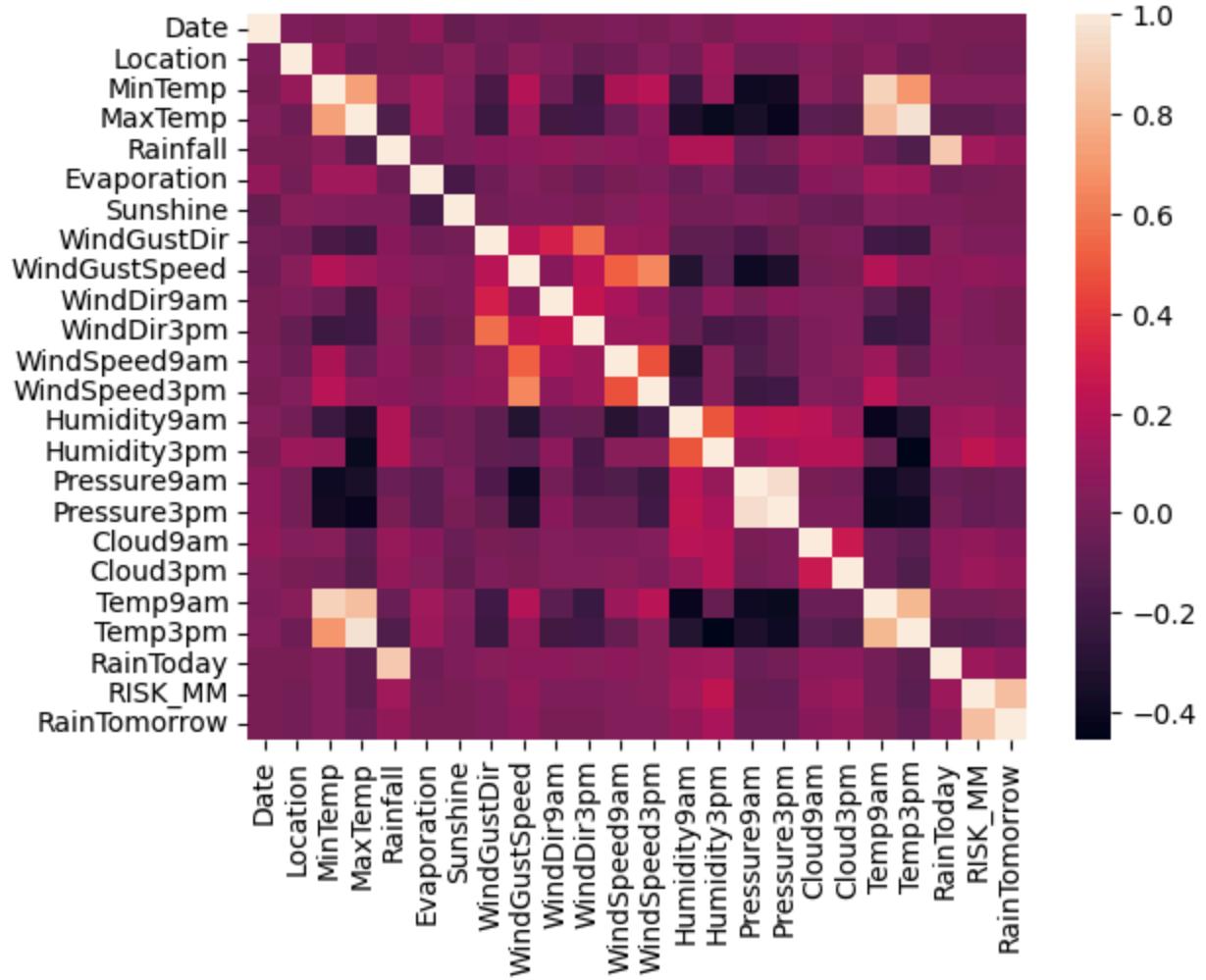
In [ ]: le=LabelEncoder()
for i in df5.columns:
    df5_clean[i]=le.fit_transform(df5_clean[i])

In [ ]: df5_clean['RainTomorrow'].unique()

Out[ ]: array([0, 1])

In [ ]: sns.heatmap(df5_clean.corr())

Out[ ]: <Axes: >
```



```
In [ ]: df5_clean
```

Out[ ]:

	Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	W
<b>0</b>	101	1	200	197	6	31	10	
<b>1</b>	102	1	139	219	0	31	10	
<b>2</b>	103	1	195	225	0	31	10	
<b>3</b>	104	1	157	248	0	31	10	
<b>4</b>	105	1	241	291	9	31	10	
...	...	...	...	...	...	...	...	...
<b>94725</b>	841	0	232	292	0	47	23	
<b>94736</b>	851	0	260	285	0	31	5	
<b>94737</b>	852	0	222	181	0	31	16	
<b>94743</b>	858	0	248	254	0	31	16	
<b>94767</b>	882	0	250	293	0	45	12	

33405 rows × 24 columns

In [ ]: `df5_clean['RainTomorrow'].unique()`

Out[ ]: `array([0, 1])`

In [ ]: `from sklearn.linear_model import LogisticRegression  
from sklearn.model_selection import train_test_split  
from sklearn.metrics import classification_report,accuracy_score`

## RAINTODAY

In [ ]: `x51=df5_clean.drop(['RainToday'],axis=1)  
y51=df5_clean['RainToday']`

In [ ]: `x51_train,x51_test,y51_train,y51_test=train_test_split(x51,y51,test_size=0.2,r`

In [ ]: `model5.fit(x51_train,y51_train)`

```
/usr/local/lib/python3.12/dist-packages/scikit-learn/linear_model/_logistic.py:465:  
ConvergenceWarning: lbfgs failed to converge (status=1):  
STOP: TOTAL NO. OF ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (`max_iter`) or scale the data as shown in:  
<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

```
Out[ ]: ▾ LogisticRegression ⓘ ?  
LogisticRegression()
```

```
In [ ]: y51_pred=model5.predict(x51_test)
```

```
In [ ]: y51_train_pred=model5.predict(x51_train)
```

```
In [ ]: accuracy_score(y51_train,y51_train_pred)
```

```
Out[ ]: 0.9971935339021104
```

```
In [ ]: accuracy_score(y51_test,y51_pred)
```

```
Out[ ]: 0.9970064361622512
```

```
In [ ]: c51=classification_report(y51_test,y51_pred)  
print(c51)
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	6303
1	0.98	0.97	0.97	378
accuracy			1.00	6681
macro avg	0.99	0.98	0.99	6681
weighted avg	1.00	1.00	1.00	6681

## RAINTOMORROW

```
In [ ]: X5 = df5_clean.drop(['RainTomorrow'], axis=1)
```

```
In [ ]: y5=df5_clean[['RainTomorrow']]
```

```
In [ ]: x5_train,x5_test,y5_train,y5_test=train_test_split(X5,y5,test_size=0.2,random_
```

```
In [ ]: x5_test
```

Out[ ]:

	Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	W
<b>59300</b>	2691	4	144	184	0	31	10	
<b>16626</b>	2036	17	170	161	0	31	10	
<b>51856</b>	1289	16	119	125	2	31	10	
<b>58025</b>	1418	4	67	99	19	31	10	
<b>16916</b>	2331	17	263	220	0	31	10	
...	...	...	...	...	...	...	...	...
<b>1173</b>	1283	1	192	231	0	31	10	
<b>49099</b>	1450	28	64	149	0	31	10	
<b>58951</b>	2341	4	136	132	0	31	10	
<b>75565</b>	2944	22	187	145	2	31	10	
<b>19648</b>	2144	19	226	221	2	31	10	

6681 rows × 23 columns

In [ ]: `model5=LogisticRegression()  
model5.fit(x5_train,y5_train)`

```
/usr/local/lib/python3.12/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().  
    y = column_or_1d(y, warn=True)  
/usr/local/lib/python3.12/dist-packages/sklearn/linear_model/_logistic.py:465:  
ConvergenceWarning: lbfsgs failed to converge (status=1):  
STOP: TOTAL NO. OF ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max\_iter) or scale the data as shown in:  
<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

`n_iter_i = _check_optimize_result(`

Out[ ]: `▼ LogisticRegression ⓘ ?`  
`LogisticRegression()`

In [ ]: `y5_pred=model5.predict(x5_test)`

In [ ]: `y5_train_pred=model5.predict(x5_train)`

In [ ]: `accuracy_score(y5_train,y5_train_pred)`

```
Out[ ]: 0.997043855710223
```

```
In [ ]: accuracy_score(y5_test,y5_pred)
```

```
Out[ ]: 0.9961083670109265
```

```
In [ ]: c5=classification_report(y5_test,y5_pred)
print(c5)
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	6407
1	0.95	0.96	0.95	274
accuracy			1.00	6681
macro avg	0.97	0.98	0.98	6681
weighted avg	1.00	1.00	1.00	6681