### In [130]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
import re
from sklearn.datasets import load_digits
from sklearn.model_selection import train_test_split
```

#### In [538]:

```
a=pd.read_csv(r"C:\Users\user\Downloads\C10_air\madrid_2013.csv")
a
```

#### Out[538]:

date	BEN	СО	EBE	NMHC	NO	NO_2	O_3	PM10	PM25	SO_2	TCH	TOL
2013- 11-01 01:00:00	NaN	0.6	NaN	NaN	135.0	74.0	NaN	NaN	NaN	7.0	NaN	NaN
2013- 11-01 01:00:00	1.5	0.5	1.3	NaN	71.0	83.0	2.0	23.0	16.0	12.0	NaN	8.3
2013- 11-01 01:00:00	3.9	NaN	2.8	NaN	49.0	70.0	NaN	NaN	NaN	NaN	NaN	9.(
2013- 11-01 01:00:00	NaN	0.5	NaN	NaN	82.0	87.0	3.0	NaN	NaN	NaN	NaN	NaN
2013- 11-01 01:00:00	NaN	NaN	NaN	NaN	242.0	111.0	2.0	NaN	NaN	12.0	NaN	NaN
2013- 03-01 00:00:00	NaN	0.4	NaN	NaN	8.0	39.0	52.0	NaN	NaN	NaN	NaN	NaN
2013- 03-01 00:00:00	NaN	0.4	NaN	NaN	1.0	11.0	NaN	6.0	NaN	2.0	NaN	NaN
2013- 03-01 00:00:00	NaN	NaN	NaN	NaN	2.0	4.0	75.0	NaN	NaN	NaN	NaN	NaN
2013- 03-01 00:00:00	NaN	NaN	NaN	NaN	2.0	11.0	52.0	NaN	NaN	NaN	NaN	NaN
2013- 03-01 00:00:00	NaN	NaN	NaN	NaN	1.0	10.0	75.0	3.0	NaN	NaN	NaN	NaN
	2013- 11-01 01:00:00  2013- 11-01 01:00:00  2013- 11-01 01:00:00  2013- 11-01 01:00:00  2013- 11-01 01:00:00  2013- 03-01 00:00:00  2013- 03-01 00:00:00  2013- 03-01 00:00:00  2013- 03-01 00:00:00  2013- 03-01 00:00:00	2013- 11-01	2013- 11-01 NaN 0.6 01:00:00	2013- 11-01	2013- 11-01	2013-   11-01   1.5   0.5   1.3   NaN   71.0   1.0   1.00:00   2013-   11-01   3.9   NaN   2.8   NaN   49.0   2013-   11-01   NaN   0.5   NaN   NaN   NaN   82.0   2013-   11-01   NaN   NaN   NaN   NaN   NaN   242.0   2013-   11-01   NaN   NaN   NaN   NaN   NaN   8.0   242.0   2013-   01:00:00   NaN   0.4   NaN   NaN   NaN   8.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   2.0   2013-   00:00:00   NaN   NaN   NaN   NaN   NaN   2.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   2.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-   03-01   NaN   NaN   NaN   NaN   NaN   NaN   1.0   2013-	2013- 11-01	2013- 11-01	2013- 11-01	2013- 11-01	2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   11-01   01:00:00   2013-   03-01   03-0	2013-   11-01   1.5   0.5   1.3   NaN   71.0   83.0   2.0   23.0   16.0   12.0   NaN   11-01   01:00:00     3.9   NaN   2.8   NaN   49.0   70.0   NaN   Na

209880 rows × 14 columns

4

#### In [539]:

#### a.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 209880 entries, 0 to 209879
Data columns (total 14 columns):
    Column
             Non-Null Count
                             Dtype
    -----
---
             -----
                             ----
             209880 non-null object
0
    date
    BEN
1
             50462 non-null
                             float64
2
    CO
             87018 non-null float64
3
    EBE
             50463 non-null float64
4
    NMHC
             25935 non-null float64
5
             209108 non-null float64
    NO
    NO_2
             209108 non-null float64
6
    0_3
             121858 non-null float64
7
8
    PM10
             104339 non-null float64
9
    PM25
             51980 non-null float64
             86970 non-null float64
10 SO_2
11
   TCH
             25935 non-null float64
12
    TOL
             50317 non-null float64
    station 209880 non-null int64
13
dtypes: float64(12), int64(1), object(1)
memory usage: 22.4+ MB
```

## In [540]:

```
b=a.fillna(value=104)
b
```

#### Out[540]:

	date	BEN	СО	EBE	NMHC	NO	NO_2	O_3	PM10	PM25	SO_2	тсн
0	2013- 11-01 01:00:00	104.0	0.6	104.0	104.0	135.0	74.0	104.0	104.0	104.0	7.0	104.0
1	2013- 11-01 01:00:00	1.5	0.5	1.3	104.0	71.0	83.0	2.0	23.0	16.0	12.0	104.0
2	2013- 11-01 01:00:00	3.9	104.0	2.8	104.0	49.0	70.0	104.0	104.0	104.0	104.0	104.0
3	2013- 11-01 01:00:00	104.0	0.5	104.0	104.0	82.0	87.0	3.0	104.0	104.0	104.0	104.0
4	2013- 11-01 01:00:00	104.0	104.0	104.0	104.0	242.0	111.0	2.0	104.0	104.0	12.0	104.0
209875	2013- 03-01 00:00:00	104.0	0.4	104.0	104.0	8.0	39.0	52.0	104.0	104.0	104.0	104.0
209876	2013- 03-01 00:00:00	104.0	0.4	104.0	104.0	1.0	11.0	104.0	6.0	104.0	2.0	104.0
209877	2013- 03-01 00:00:00	104.0	104.0	104.0	104.0	2.0	4.0	75.0	104.0	104.0	104.0	104.0
209878	2013- 03-01 00:00:00	104.0	104.0	104.0	104.0	2.0	11.0	52.0	104.0	104.0	104.0	104.0
209879	2013- 03-01 00:00:00	104.0	104.0	104.0	104.0	1.0	10.0	75.0	3.0	104.0	104.0	104.0

209880 rows × 14 columns

## In [541]:

b.columns

## Out[541]:

# In [542]:

c=b.head(10)
c

# Out[542]:

	date	BEN	со	EBE	NMHC	NO	NO_2	O_3	PM10	PM25	SO_2	тсн	TC
0	2013- 11-01 01:00:00	104.0	0.6	104.0	104.00	135.0	74.0	104.0	104.0	104.0	7.0	104.00	104
1	2013- 11-01 01:00:00	1.5	0.5	1.3	104.00	71.0	83.0	2.0	23.0	16.0	12.0	104.00	8
2	2013- 11-01 01:00:00	3.9	104.0	2.8	104.00	49.0	70.0	104.0	104.0	104.0	104.0	104.00	9
3	2013- 11-01 01:00:00	104.0	0.5	104.0	104.00	82.0	87.0	3.0	104.0	104.0	104.0	104.00	104
4	2013- 11-01 01:00:00	104.0	104.0	104.0	104.00	242.0	111.0	2.0	104.0	104.0	12.0	104.00	104
5	2013- 11-01 01:00:00	1.0	0.6	0.8	104.00	70.0	70.0	2.0	24.0	104.0	6.0	104.00	5
6	2013- 11-01 01:00:00	104.0	0.4	104.0	0.29	51.0	80.0	5.0	23.0	14.0	4.0	1.44	104
7	2013- 11-01 01:00:00	104.0	104.0	104.0	0.23	29.0	60.0	4.0	104.0	104.0	104.0	1.51	104
8	2013- 11-01 01:00:00	104.0	1.0	104.0	104.00	165.0	107.0	2.0	104.0	104.0	11.0	104.00	104
9	2013- 11-01 01:00:00	104.0	0.6	104.0	104.00	63.0	93.0	104.0	11.0	104.0	8.0	104.00	104
4.4													

## In [543]:

```
d=c[['BEN', 'CO', 'EBE', 'NMHC', 'NO_2', 'O_3',
    'PM10', 'SO_2', 'TCH', 'TOL', 'station']]
d
```

## Out[543]:

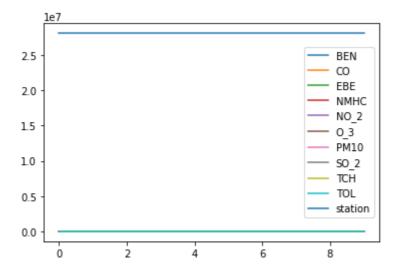
	BEN	СО	EBE	NMHC	NO_2	O_3	PM10	SO_2	тсн	TOL	station
0	104.0	0.6	104.0	104.00	74.0	104.0	104.0	7.0	104.00	104.0	28079004
1	1.5	0.5	1.3	104.00	83.0	2.0	23.0	12.0	104.00	8.3	28079008
2	3.9	104.0	2.8	104.00	70.0	104.0	104.0	104.0	104.00	9.0	28079011
3	104.0	0.5	104.0	104.00	87.0	3.0	104.0	104.0	104.00	104.0	28079016
4	104.0	104.0	104.0	104.00	111.0	2.0	104.0	12.0	104.00	104.0	28079017
5	1.0	0.6	8.0	104.00	70.0	2.0	24.0	6.0	104.00	5.2	28079018
6	104.0	0.4	104.0	0.29	80.0	5.0	23.0	4.0	1.44	104.0	28079024
7	104.0	104.0	104.0	0.23	60.0	4.0	104.0	104.0	1.51	104.0	28079027
8	104.0	1.0	104.0	104.00	107.0	2.0	104.0	11.0	104.00	104.0	28079035
9	104.0	0.6	104.0	104.00	93.0	104.0	11.0	8.0	104.00	104.0	28079036

## In [544]:

```
d.plot.line()
```

#### Out[544]:

## <AxesSubplot:>

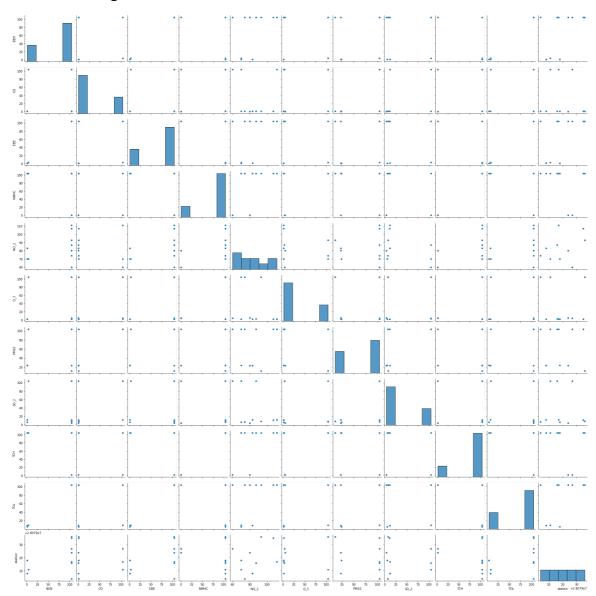


## In [545]:

```
sns.pairplot(d)
```

## Out[545]:

<seaborn.axisgrid.PairGrid at 0x1186e23bfa0>



#### In [546]:

```
x=d[['BEN', 'CO', 'EBE', 'NMHC', 'NO_2']]
y=d['TCH']
```

## In [547]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

#### In [548]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

#### Out[548]:

LinearRegression()

## In [549]:

```
print(lr.intercept_)
```

#### 1.1532157632747015

#### In [550]:

```
coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

## Out[550]:

#### Co-efficient

BEN -3.339620e-12

CO 2.854507e-14

**EBE** 3.332956e-12

**NMHC** 9.889114e-01

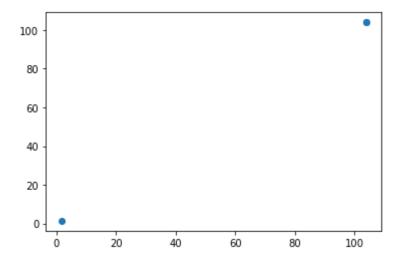
**NO\_2** -1.138840e-16

#### In [551]:

```
prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

#### Out[551]:

<matplotlib.collections.PathCollection at 0x118772352e0>



```
In [552]:
print(lr.score(x_test,y_test))
0.999997611321236
In [553]:
from sklearn.linear_model import Ridge,Lasso
In [554]:
rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
Out[554]:
Ridge(alpha=10)
In [555]:
rr.score(x_test,y_test)
Out[555]:
0.9999995125979597
In [556]:
la=Lasso(alpha=10)
la.fit(x_train,y_train)
Out[556]:
Lasso(alpha=10)
In [557]:
la.score(x_test,y_test)
Out[557]:
```

0.9999538023684293

## In [558]:

```
a1=b.head(7000)
a1
```

## Out[558]:

	date	BEN	со	EBE	NMHC	NO	NO_2	O_3	PM10	PM25	SO_2	тсн	
0	2013- 11-01 01:00:00	104.0	0.6	104.0	104.0	135.0	74.0	104.0	104.0	104.0	7.0	104.0	1
1	2013- 11-01 01:00:00	1.5	0.5	1.3	104.0	71.0	83.0	2.0	23.0	16.0	12.0	104.0	
2	2013- 11-01 01:00:00	3.9	104.0	2.8	104.0	49.0	70.0	104.0	104.0	104.0	104.0	104.0	
3	2013- 11-01 01:00:00	104.0	0.5	104.0	104.0	82.0	87.0	3.0	104.0	104.0	104.0	104.0	1
4	2013- 11-01 01:00:00	104.0	104.0	104.0	104.0	242.0	111.0	2.0	104.0	104.0	12.0	104.0	1
6995	2013- 11-13 04:00:00	104.0	0.2	104.0	104.0	1.0	8.0	40.0	104.0	104.0	104.0	104.0	1
6996	2013- 11-13 04:00:00	104.0	104.0	104.0	104.0	1.0	5.0	104.0	3.0	104.0	1.0	104.0	1
6997	2013- 11-13 04:00:00	104.0	104.0	104.0	104.0	1.0	6.0	104.0	3.0	2.0	104.0	104.0	1
6998	2013- 11-13 04:00:00	104.0	104.0	104.0	104.0	1.0	9.0	104.0	5.0	1.0	104.0	104.0	1
6999	2013- 11-13 04:00:00	104.0	104.0	104.0	104.0	1.0	9.0	43.0	104.0	104.0	104.0	104.0	1

#### 7000 rows × 14 columns

# In [559]:

```
e=a1[['BEN', 'CO', 'EBE', 'NMHC', 'NO_2', 'O_3', 'PM10', 'SO_2', 'TCH', 'TOL', 'station']]
```

# In [560]:

```
f=e.iloc[:,0:14]
g=e.iloc[:,-1]
```

```
In [561]:
h=StandardScaler().fit_transform(f)
In [562]:
logr=LogisticRegression(max_iter=10000)
logr.fit(h,g)
Out[562]:
LogisticRegression(max_iter=10000)
In [563]:
from sklearn.model_selection import train_test_split
h_train,h_test,g_train,g_test=train_test_split(h,g,test_size=0.3)
In [564]:
i=[[10,20,30,40,50,60,11,22,33,44,55]]
In [565]:
prediction=logr.predict(i)
print(prediction)
[28079050]
In [566]:
logr.classes_
Out[566]:
array([28079004, 28079008, 28079011, 28079016, 28079017, 28079018,
       28079024, 28079027, 28079035, 28079036, 28079038, 28079039,
       28079040, 28079047, 28079048, 28079049, 28079050, 28079054,
       28079055, 28079056, 28079057, 28079058, 28079059, 28079060],
      dtype=int64)
In [567]:
logr.predict_proba(i)[0][0]
Out[567]:
0.0
In [568]:
logr.predict_proba(i)[0][1]
Out[568]:
0.0
```

```
In [569]:
logr.score(h_test,g_test)
Out[569]:
0.9552380952380952
In [570]:
from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
Out[570]:
ElasticNet()
In [571]:
print(en.coef_)
             0.
                        -0.
                                    0.9881566 0.
[-0.
In [572]:
print(en.intercept_)
1.2205308629257416
In [573]:
prediction=en.predict(x_test)
print(en.score(x_test,y_test))
0.9999994119369457
In [574]:
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(h_train,g_train)
Out[574]:
RandomForestClassifier()
In [575]:
parameters={'max_depth':[1,2,3,4,5],
 'min_samples_leaf':[5,10,15,20,25],
 'n_estimators':[10,20,30,40,50]
 }
```

```
In [576]:
```

```
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="accuracy")
grid_search.fit(h_train,g_train)
```

## Out[576]:

#### In [577]:

```
grid_search.best_score_
```

## Out[577]:

0.9979591836734694

#### In [578]:

rfc\_best=grid\_search.best\_estimator\_

# In [579]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,50))
plot_tree(rfc_best.estimators_[2],filled=True)
```

Out[579]:

```
[\text{Text}(1879.578947368421, 2491.5, 'X[10] <= 1.236 \text{ ngini} = 0.958 \text{ nsamples} =
209, 190, 192, 191, 221, 199, 193, 212\n185, 232, 194, 219]'),
 Text(1644.6315789473683, 20 <= -1.109\ngini = 0.956\nsamples =
2898\nvalue = [215, 192, 212, 222, 205, 201, 208, 195, 218, 205\n187, 203,
209, 190, 192, 191, 221, 199, 193, 212\n185, 232, 194, 0]'),
 Text(704.8421052631579, 1585 5 'X[6] <= 0.456\ngini = 0.666\nsamples = 3
70\nvalue = [0, 0, 0, 0, 0, 0, 0, 195, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 195]
3, 0, 0, 0, 0, 0]'),
 Text(469.89473684210526, 1132.5, 'X[5] <= 0.716\ngini = 0.499\nsamples =
251\nvalu______, 0, 0, 0, 0, 0, 207, 0, _______, 0, 0, 0\n0, 0, 0, 0, 193,
0, 0, 0, 0, 0]'),
 Text(234,94736842105263, 679.5, 'gini = 0.0\nsamples = 127\nvalue = [0,
Text(704.8421052631579, 679.5, 'gini = 0.0\nsamples = 124\nvalue = [0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0, 0, 193, 0, 0, 0, 0]'),
Text(2584.4210526315787, 1585.5, X[1] <= -0.142 \setminus gini = 0.95 \setminus gini =
2528\nvalue = [215, 192, 212, 222, 205, 201, 1, 0, 218, 205, 187\n203, 20
Text(1644.6315789473683, 1132.5, 'X[5] <= 0.604\ngini = 0.889\nsamples =
1146\nvalue = [215, 192, 0, 222, 0, 201, 1, 0, 218, 205, 0, 203\n0, 0, 0,
0, 0, 0, 0, 212, 185, 0, 0, 0]'),
Ime\t(1174.7368421052631, 679.5, 'X[10] <= 0.555\ngini = 0.833\nsamples =
0, 0, 212, 0, 0, 0, 0]'),
 Text(939.7894736842105, 226.5, 'gini = 0.8\nsamples = 635\nvalue = [0, 19
2, 0, 222, 0, 201, 1, 0, 218, 0, 0, 203\n0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0]'),
 Text(1409.6842105263158, 226.5, 'gini = 0.0\nsamples = 140\nvalue = [0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0, 0, 0, 212, 0, 0, 0]'),
 Text(2114.5263157894738, 679.5, 'X[6] \leftarrow 0.445  | mgini = 0.665 | nsamples = 3
71\nvalue = [215, 0, 0, 0, 0, 0, 0, 0, 205, 0, 0, 0, 0 \n0, 0, 0, 0, 0, 0]
a 185 a a al'\
```