```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn import preprocessing
label encoder = preprocessing.LabelEncoder()
# Encode labels in column 'species'.
# Importing the dataset
df = pd.read csv('Desktop/pollution.csv')
df['Air Quality']= label encoder.fit transform(df['Air Quality'])
X=df.iloc[:,:-1]
y=df.iloc[:,-1]
```

```
location month year SO2 μg/l NO2μg/l PM10 μg/l \
0 CLOCK TOWER-DEHRADUN
                         1 2012
                                   27.33
                                           30.33
                                                    193.28
  CLOCK TOWER-DEHRADUN
                         2 2012 25.68 25.80
                                                    173.77
                                           27.50
2 CLOCK TOWER-DEHRADUN
                         3 2012
                                  29.64
                                                    211.35
  PM2.5 μ g/l CO μg/l O3 μ g/l 8 HR NH3 μ g/l
                                               AOI
        60.0
                              100
                                         400 162.19
0
                   2
п
        60.0
                   2
                              100
                                         400 149.18
2
        60.0
                              100
                                         400 174.23
```

```
from sklearn.preprocessing import OneHotEncoder
enc = OneHotEncoder()
# transforming the column after fitting
enc = enc.fit_transform(X[['location']]).toarray()
# converting arrays to a dataframe
encoded_colm = pd.DataFrame(enc)
# concating dataframes
X = pd.concat([X, encoded_colm], axis = 1)
# removing the encoded column.
X = X.drop(['location'], axis = 1)
```

D

X.head(3)

```
X.head(5)
              502 \mu g/1 N02 \mu g/1 PM10 \mu g/1 PM2.5 \mu g/1 CO \mu g/1 \
  month year
0
      1 2012
                  27.33
                          30.33
                                    193.28
                                                  60.0
                                                             2
      2 2012
                  25.68
                          25.80
                                    173.77
                                                  60.0
                                                             2
2
      3 2012
                  29.64
                          27.50
                                    211.35
                                                  60.0
                                                             2
3
      4 2012
                  28.64
                          26.81
                                    230.76
                                                  60.0
                                                             2
4
      5 2012
                  31.09
                          29.30
                                    310.73
                                                  60.0
                                                             2
                                                    3
  03 μg/1 8 HR NH3 μg/1
                               AQI
                                      0
                                          1
                                               2
                                                         4
                                                             5
                                                                  6
0
            100
                       400
                            162.19 1.0 0.0
                                             0.0
                                                  0.0
                                                       0.0
                                                           0.0
                                                                0.0 0.0
            100
                       400
                            149.18
                                   1.0
                                        0.0
                                             0.0
                                                  0.0
                                                       0.0
                                                           0.0
                                                                0.0 0.0
2
            100
                            174.23
                                   1.0
                                        0.0
                                             0.0
                                                  0.0
                                                       0.0
                                                           0.0
                       400
                                                                0.0 0.0
3
            100
                       400
                            187.17 1.0
                                        0.0
                                             0.0 0.0
                                                      0.0
                                                           0.0
                                                                0.0 0.0
4
            100
                            260.73 1.0
                                        0.0
                                             0.0
                                                  0.0
                                                       0.0
                                                           0.0
                       400
                                                                0.0 0.0
```

```
y.head(5)
    0
    θ
    0
    0
    1
Name: Air Quality, dtype: int32
   # Splitting the dataset into the Training set and Test set
   from sklearn.model selection import train test split
   X train, X test, y train, y test = train_test split(X, y, test_size = 0.20, random_state = 0)
   # Feature Scaling
   from sklearn.preprocessing import StandardScaler
   sc = StandardScaler()
   X train = sc.fit transform(X train)
   X_test = sc.transform(X_test)
```

```
# Feature Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X train = sc.fit transform(X train)
X test = sc.transform(X test)
# Fitting Decision Tree Classification to the Training set
from sklearn.tree import DecisionTreeClassifier
```

```
# Fitting Decision Tree Classification to the Training set
from sklearn.tree import DecisionTreeClassifier
classifier = DecisionTreeClassifier(criterion = 'entropy', random_state = θ)
classifier.fit(X_train, y_train)
```

from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy score

Predicting the Test set results
y_pred = classifier.predict(X_test)

```
#printing the accuracy of Decision tree
accuracy_score(y_test, y_pred)

0.9895833333333334

# Fitting SVM to the Training set
from sklearn.svm import SVC
```

classifier.fit(X train, y train)

Predicting the Test set results
y pred = classifier.predict(X test)

accuracy score(y test, y pred)

#Accuracy of SVM

0.97395833333333334

classifier = SVC(kernel = 'linear', random_state = 0)

```
0.97395833333333334
```

```
#fitting knn model
from sklearn.neighbors import KNeighborsClassifier
classifier=KNeighborsClassifier(n_neighbors=5,metric='minkowski',p=2)
classifier.fit(X_train,y_train)

# Predicting the Test set results
y_pred = classifier.predict(X_test)
```

```
# Predicting the Test set results
y_pred = classifier.predict(X_test)

# Making the Confusion Matrix
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
```

```
#Accuracy of knn
accuracy_score(y_test, y_pred)
```