```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

Reading the Data

```
mData = pd.read_csv('weatherAUS.csv')
print('The Shape of The Data ',mData.shape)
mData.info()
```



```
The Shape of The Data (142193, 24)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 142193 entries, 0 to 142192
Data columns (total 24 columns):
Date 142193 non-null object
Location 142193 non-null object
```

Preprocessing the Data

```
from sklearn import preprocessing
from sklearn.utils import resample
    MTIIODTT 2 OIII
                      TOTION HOH-HATT ON LECT
def Preprocessing Data(data):
  mDataDroped = data.drop(columns=['Date','RISK MM','Sunshine','Evaporation','Cloud3pm','Cloud9am','Pressure3pm
                        ,axis=1)
  mDataDroped = mDataDroped.dropna(how='any')
 mDataDroped= mDataDroped.reset index()
  Y = mDataDroped['RainTomorrow']
  X = mDataDroped.drop(columns=['RainTomorrow'])
  col names = X.select dtypes("object").columns
  X = pd.get dummies(X, columns=col names)
  scaler = preprocessing.MinMaxScaler().fit(X)
  x normalized = scaler.transform(X)
  newData = pd.DataFrame(columns= X.columns,data=x normalized)
  newData['RainTomorrow'] = Y.map({'Yes': 1, 'No': 0})
  # -----Ballancing the Data set------
  #Categories = pd.Categorical(Y.astype('object')).categories
  #numOfCategories = Categories.size
  #CategorySizes = Y.value counts()
  #maxSize = Y.value counts().max()
  #mSampledDate =[]
  #for Category in Categories :
   if (newData[Y==Category].shape[0] != maxSize):
      temp = resample(newData[Y==Category],
                                        # sample with replacement
                      replace=True,
                      n samples=(maxSize-data[Y==Category].shape[0]),
                      random state=123) # reproducible results
      mSampledDate.append(temp)
 #BallancedData=newData
  #for Data in mSampledDate:
 # BallancedData = pd.concat([BallancedData, Data])
  #print("total shape after up sampling = ", BallancedData.shape)
```

```
#BallancedData = pd.DataFrame(BallancedData)
#BallancedData.index = np.array(range(0,BallancedData.shape[0]))
return newData , scaler
```

DataFinal,scaler = Preprocessing_Data(mData)
DataFinal



index	MinTemp	MaxTemp	Rainfall	WindGustSpeed	WindSpeed9am	WindSpeed3pm	Humidity9am	Humidity3pm	Temp
0.000000	0.516509	0.523629	0.001632	0.289062	0.211765	0.258824	0.71	0.22	0.508
0.000007	0.375000	0.565217	0.000000	0.289062	0.023529	0.235294	0.44	0.25	0.514
0.000014	0.504717	0.576560	0.000000	0.304688	0.200000	0.282353	0.38	0.30	0.594
0.000021	0.417453	0.620038	0.000000	0.132812	0.105882	0.082353	0.45	0.16	0.533
0.000028	0.613208	0.701323	0.002720	0.265625	0.058824	0.211765	0.82	0.33	0.527
0.000035	0.544811	0.652174	0.000544	0.382812	0.200000	0.258824	0.55	0.23	0.586
0.000042	0.537736	0.563327	0.000000	0.335938	0.211765	0.258824	0.49	0.19	0.533
0.000049	0.382075	0.595463	0.000000	0.218750	0.047059	0.176471	0.48	0.19	0.495
0.000056	0.429245	0.693762	0.000000	0.570312	0.058824	0.305882	0.42	0.09	0.537
0.000063	0.509434	0.659735	0.003808	0.164062	0.152941	0.105882	0.58	0.27	0.575
0.000070	0.516509	0.665406	0.000000	0.179688	0.176471	0.047059	0.48	0.22	0.582
0.000077	0.575472	0.500945	0.005985	0.187500	0.152941	0.129412	0.89	0.91	0.487
0.000084	0.575472	0.442344	0.042437	0.421875	0.305882	0.305882	0.76	0.93	0.518
0.000091	0.497642	0.487713	0.009793	0.289062	0.258824	0.211765	0.65	0.43	0.485
0.000105	0.533019	0.485822	0.000000	0.117188	0.105882	0.082353	0.69	0.82	0.514
0.000113	0.518868	0.523629	0.045702	0.437500	0.047059	0.211765	0.80	0.65	0.531
0.000120	0.464623	0.516068	0.028836	0.281250	0.258824	0.176471	0.47	0.32	0.478
0.000127	0.431604	0.574669	0.000000	0.148438	0.176471	0.047059	0.45	0.26	0.485
0.000134	0.471698	0.644612	0.000000	0.132812	0.082353	0.082353	0.56	0.28	0.554
0.000141	0.603774	0.714556	0.000000	0.281250	0.176471	0.235294	0.38	0.28	0.668
0.000148	0.683962	0.691871	0.000000	0.265625	0.200000	0.211765	0.54	0.24	0.654
0.000155	0.561321	0.674858	0.000000	0.203125	0.047059	0.129412	0.55	0.23	0.592
	0.000000 0.000007 0.000014 0.000021 0.000028 0.000035 0.000042 0.000056 0.000056 0.000077 0.000077 0.000084 0.000091 0.000105 0.000113 0.000127 0.000127 0.000134 0.000141	0.0000070.3750000.00000140.5047170.0000210.4174530.0000280.6132080.0000350.5448110.0000420.5377360.0000560.4292450.0000630.5094340.0000700.5165090.0000770.5754720.0000840.5754720.0000910.4976420.0001130.5188680.0001200.4646230.0001340.4716980.0001410.6037740.0001480.683962	0.0000000.5165090.5236290.0000070.3750000.5652170.0000140.5047170.5765600.0000210.4174530.6200380.0000280.6132080.7013230.0000350.5448110.6521740.0000420.5377360.5633270.0000490.3820750.5954630.0000560.4292450.6937620.0000700.5165090.6654060.0000770.5754720.5009450.0000840.5754720.4423440.0000910.4976420.4877130.0001050.5330190.4858220.0001130.5188680.5236290.0001270.4316040.5746690.0001340.4716980.6446120.0001410.6037740.714556	0.000000 0.516509 0.523629 0.001632 0.000007 0.375000 0.565217 0.000000 0.000014 0.504717 0.576560 0.000000 0.000021 0.417453 0.620038 0.0002720 0.000035 0.544811 0.652174 0.000544 0.000042 0.537736 0.563327 0.000000 0.000049 0.382075 0.595463 0.000000 0.000056 0.429245 0.693762 0.0003808 0.000070 0.516509 0.665406 0.000000 0.000077 0.575472 0.500945 0.005985 0.000084 0.575472 0.442344 0.042437 0.000091 0.497642 0.487713 0.009793 0.000105 0.533019 0.485822 0.000000 0.000120 0.464623 0.516068 0.028836 0.000127 0.431604 0.574669 0.000000 0.000134 0.471698 0.644612 0.000000 0.000148 0.683962 0	0.000000 0.516509 0.523629 0.001632 0.289062 0.000007 0.375000 0.565217 0.000000 0.389062 0.000014 0.504717 0.576560 0.000000 0.304688 0.000028 0.613208 0.701323 0.002720 0.265625 0.000035 0.544811 0.652174 0.000544 0.382812 0.000042 0.537736 0.563327 0.000000 0.335938 0.000049 0.382075 0.595463 0.000000 0.570312 0.000056 0.429245 0.693762 0.000000 0.570312 0.000070 0.516509 0.665406 0.000000 0.179688 0.000077 0.575472 0.500945 0.005985 0.187500 0.000084 0.575472 0.442344 0.042437 0.421875 0.000105 0.533019 0.485822 0.000000 0.117188 0.000113 0.518868 0.523629 0.045702 0.437500 0.000127 0.431604 0.574669 0.000000	0.000000 0.516509 0.523629 0.001632 0.289062 0.211765 0.000007 0.375000 0.565217 0.000000 0.289062 0.023529 0.000014 0.504717 0.576560 0.000000 0.304688 0.200000 0.000021 0.417453 0.620038 0.000000 0.132812 0.105882 0.000028 0.613208 0.701323 0.002720 0.265625 0.058824 0.000035 0.544811 0.652174 0.000544 0.382812 0.200000 0.000042 0.537736 0.563327 0.000000 0.335938 0.211765 0.000049 0.382075 0.595463 0.000000 0.570312 0.058824 0.000056 0.429245 0.693762 0.000000 0.570312 0.058824 0.000070 0.516509 0.665406 0.0003808 0.164062 0.152941 0.000077 0.575472 0.500945 0.005985 0.187500 0.152941 0.000084 0.575472 0.442344 0.042437	0.000000 0.516509 0.523629 0.001632 0.289062 0.211765 0.258824 0.000007 0.375000 0.565217 0.000000 0.289062 0.023529 0.235294 0.000014 0.504717 0.576560 0.000000 0.304688 0.200000 0.282353 0.000021 0.417453 0.620038 0.000000 0.132812 0.105882 0.082353 0.000028 0.613208 0.701323 0.002720 0.265625 0.058824 0.211765 0.000035 0.544811 0.652174 0.000544 0.382812 0.200000 0.258824 0.000042 0.537736 0.563327 0.000000 0.335938 0.211765 0.258824 0.000049 0.382075 0.595463 0.000000 0.570312 0.058824 0.305882 0.000060 0.429245 0.693762 0.003808 0.164062 0.152941 0.105882 0.000070 0.575472 0.500945 0.005985 0.187500 0.152941 0.129412 0.000105	0.000000 0.516509 0.523629 0.001632 0.289062 0.211765 0.258824 0.71 0.000007 0.375000 0.565217 0.000000 0.289062 0.023529 0.235294 0.44 0.000014 0.504717 0.576560 0.000000 0.304688 0.200000 0.282353 0.38 0.000021 0.417453 0.620038 0.000000 0.132812 0.105882 0.082353 0.45 0.000028 0.613208 0.701323 0.002720 0.265625 0.058824 0.211765 0.82 0.000042 0.537736 0.563327 0.000000 0.335938 0.211765 0.258824 0.55 0.000042 0.537736 0.595463 0.000000 0.218750 0.047059 0.176471 0.48 0.000060 0.429245 0.693762 0.000000 0.570312 0.058824 0.305882 0.42 0.000070 0.516509 0.665406 0.000000 0.179688 0.176471 0.047059 0.48 0.000072	0.000000 0.516509 0.523629 0.001632 0.289062 0.211765 0.258824 0.71 0.22 0.000007 0.375000 0.565217 0.000000 0.289062 0.023529 0.235294 0.44 0.25 0.000014 0.504717 0.576560 0.000000 0.304688 0.200000 0.282353 0.38 0.30 0.000021 0.417453 0.620038 0.000000 0.132812 0.105882 0.082353 0.45 0.16 0.000028 0.613208 0.701323 0.002720 0.265625 0.058824 0.211765 0.82 0.33 0.000042 0.537736 0.563327 0.000000 0.335938 0.211765 0.258824 0.49 0.19 0.000049 0.382075 0.595463 0.000000 0.218750 0.047059 0.176471 0.48 0.19 0.000060 0.429245 0.693762 0.000000 0.570312 0.058824 0.305882 0.42 0.09 0.000070 0.516509 0.665406

22	0.000162	0.497642	0.703214	0.000000	0.281250	0.023529	0.200000	0.49	0.17	0.605
23	0.000169	0.582547	0.731569	0.000000	0.218750	0.082353	0.129412	0.45	0.19	0.641
24	0.000183	0.674528	0.708885	0.000000	0.320312	0.129412	0.329412	0.56	0.15	0.670
25	0.000190	0.665094	0.604915	0.000000	0.304688	0.200000	0.329412	0.49	0.22	0.607
26	0.000197	0.495283	0.548204	0.003264	0.335938	0.105882	0.235294	0.78	0.70	0.415
27	0.000204	0.483491	0.551985	0.002176	0.250000	0.176471	0.176471	0.48	0.28	0.508
28	0.000211	0.466981	0.591682	0.000000	0.382812	0.200000	0.341176	0.46	0.26	0.567
29	0.000218	0.426887	0.542533	0.000000	0.265625	0.200000	0.105882	0.44	0.22	0.466
121760	0.999789	0.544811	0.587902	0.000000	0.234375	0.200000	0.211765	0.61	0.36	0.476
121761	0.999803	0.419811	0.620038	0.000000	0.164062	0.047059	0.105882	0.36	0.16	0.502
121762	0.999810	0.389151	0.555766	0.000000	0.203125	0.105882	0.129412	0.46	0.25	0.443
121763	0.999817	0.500000	0.510397	0.000000	0.234375	0.200000	0.129412	0.59	0.34	0.445
121764	0.999824	0.422170	0.519849	0.000000	0.218750	0.129412	0.176471	0.62	0.32	0.400
121765	0.999831	0.327830	0.478261	0.000000	0.304688	0.211765	0.305882	0.56	0.32	0.386
121766	0.999838	0.332547	0.457467	0.000000	0.281250	0.235294	0.258824	0.61	0.22	0.356
121767	0.999845	0.235849	0.453686	0.000000	0.234375	0.176471	0.235294	0.45	0.18	0.295
121768	0.999852	0.290094	0.465028	0.000000	0.250000	0.200000	0.176471	0.42	0.22	0.333
121769	0.999859	0.306604	0.446125	0.000000	0.187500	0.129412	0.152941	0.42	0.26	0.352
121770	0.999866	0.316038	0.482042	0.000000	0.234375	0.176471	0.258824	0.38	0.11	0.369
121771	0.999873	0.228774	0.483932	0.000000	0.281250	0.129412	0.305882	0.29	0.06	0.318
121772	0.999880	0.212264	0.500945	0.000000	0.281250	0.129412	0.211765	0.27	0.19	0.352
121773	0.999887	0.294811	0.468809	0.000000	0.250000	0.129412	0.282353	0.58	0.26	0.343

Data Helping Functions

```
def GetFeature_Output(data):
    y = data['RainTomorrow']
    x = data.drop(columns=['RainTomorrow'])
    return x,y

def addOutput(x,y):
    x["RainTomorrow"]=y
    return x
```

Results Variables

```
myResultsDF = pd.DataFrame()
myResultsDF['Metric'] = ['completeness_score', 'adjusted_rand_score', 'fowlkes_mallows_score', 'silhouette_score', 'calinski_hara
myResultsDF
```



Metric

- 0 completeness_score
- 1 adjusted_rand_score
- 2 fowlkes_mallows_score
- 3 silhouette_score
- 4 calinski_harabaz_score
- 5 CentroidRMSE

Evaluation Functions

```
#Centroid Evaluation Function
def CentroidRMSE(Features, Clutered1, Clustered2):
```

```
#print("C1 IN ",Clutered1.shape)
    #print("C2 IN ",Clustered2.shape)
    X1 = Features.copv()
    X1["C1"] = Clutered1
    C1Means = X1.groupby("C1").mean().sort values(X1.columns[0])
    X2 = Features.copy()
    X2["C2"] = Clustered2
    C2Means = X2.groupby("C2").mean().sort values(X2.columns[0])
   #print("C1 IN Arranged ",C1Means.shape)
    #print("C2 IN Arranged", C2Means.shape)
    #if(C1Means.shape[0]>C2Means.shape[0]):
    # diff = C2Means.shape[0] - C1Means.shape[0]
    # C1Means = C1Means[:diff]
    # print("C1 IN Changed", C1Means.shape)
    #elif(C1Means.shape[0]<C2Means.shape[0]):</pre>
    # diff = C1Means.shape[0] - C2Means.shape[0]
    # C2Means = C2Means[:diff]
    # print("C2 IN Changed ",C2Means.shape)
    RMSE = np.sqrt(metrics.mean squared error(C1Means,C2Means))
    return RMSE
def Evalution(X,Y,Y Pred):
  result = []
  result.append(metrics.completeness score(Y, Y Pred))
  result.append(metrics.adjusted rand score(Y, Y Pred))
  result.append(metrics.fowlkes mallows score(Y,Y Pred))
  result.append(metrics.silhouette score(X, Y Pred, metric='euclidean'))
  result.append(metrics.calinski harabaz score(X, Y Pred))
  result.append(CentroidRMSE(X,Y,Y Pred))
  return result
```

K_Mean Test

```
from sklearn.cluster import KMeans
from sklearn import metrics
from sklearn.model_selection import train_test_split,StratifiedKFold

def K_Mean_Test(data,clusters_num):
    # full Data
```

```
X , Y = GetFeature Output(data)
  Clusterer = KMeans(n clusters=clusters num, random state=0).fit(X)
  Y Result = Clusterer.predict(X)
  Test Result = Evalution(X,Y.values,Y Result)
  # Train Test
  x train, x test, y train, y test = train test split(X, Y, test size=0.4, random state=0)
  Clusterer = KMeans(n clusters=clusters num, random state=0).fit(x train)
  Y Result = Clusterer.predict(x test)
  result train test = Evalution(x test, y test, Y Result)
  KfordScorer = []
  k fold = StratifiedKFold(n splits=5)
  Clusterer = KMeans(n clusters=clusters num, random state=0)
  for train indices, test indices in k fold.split(X,Y):
    Clusterer.fit(X.iloc[train indices])
    Y Result = Clusterer.predict(X.iloc[test indices])
    x test = X.iloc[test indices]
    y test = Y.iloc[test indices]
    resultTemp = Evalution(x test, y test.values, Y Result)
    KfordScorer.append(resultTemp)
  results KFolds= np.mean(KfordScorer,axis=0)
  return Test Result, result train test, results KFolds
r1,r2,r3 = K Mean Test(DataFinal,2)
myResultsDF['Kmean Full'] = r1
myResultsDF['Kmean FTrain Test'] = r2
myResultsDF['Kmean FK-folds'] = r3
myResultsDF
```

	Metric	Kmean_Full	Kmean_FTrain_Test	Kmean_FK-folds
0	completeness score	0.082771	0.083151	0.084843

PCA -> K_Mean Test

```
3
              silhouette score
                                 0 109456
                                                      0 109493
                                                                       0 115697
from sklearn.decomposition import PCA
def PCA K Mean Test(data, clusters num, PCA comp num):
 # full Data
 X Original, Y Original = GetFeature Output(data)
  scikit pca = PCA(n components=PCA comp num)
 X = scikit pca.fit transform(X Original)
  X = pd.DataFrame(X)
  Clusterer = KMeans(n clusters=clusters num, random state=0).fit(X)
 Y Result = Clusterer.predict(X)
  Test Result = Evalution(X Original, Y Original.values, Y Result)
  # Train Test
  x train, x test, y train, y test = train test split(X, Y Original, test size=0.4, random state=0)
  Clusterer = KMeans(n clusters=clusters num, random state=0).fit(x train)
  Y Result = Clusterer.predict(x test)
  result train test = Evalution(X Original.iloc[x test.index], y test, Y Result)
  KfordScorer = []
  k fold = StratifiedKFold(n splits=5)
  Clusterer = KMeans(n clusters=clusters num, random state=0)
  for train indices, test indices in k fold.split(X,Y Original):
    Clusterer.fit(X.iloc[train indices])
    Y Result = Clusterer.predict(X.iloc[test indices])
    x test = X.iloc[test indices]
    y test = Y Original.iloc[test indices]
    resultTemp = Evalution(X Original.iloc[x test.index], y test, Y Result)
    KfordScorer.append(resultTemp)
  results KFolds= np.mean(KfordScorer,axis=0)
  return Test Result, result train test, results KFolds
```

```
r1,r2,r3 = PCA_K_Mean_Test(DataFinal,2,50)

myResultsDF['PCA_Full'] = r1
myResultsDF['PCA_Train_Test'] = r2
myResultsDF['PCA_K-folds'] = r3
myResultsDF
```

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	Metric	Kmean_Full	Kmean_FTrain_Test	Kmean_FK-folds	PCA_Full	PCA_Train_Test	PCA_K-folds
0	completeness_score	0.082771	0.083151	0.084843	0.082771	0.083151	0.084843
1	adjusted_rand_score	0.199819	0.200140	0.201305	0.199819	0.200140	0.201305
2	fowlkes_mallows_score	0.723365	0.723724	0.723783	0.723365	0.723724	0.723783
3	silhouette_score	0.109456	0.109493	0.115697	0.109456	0.109493	0.115697
4	calinski_harabaz_score	11087.223740	4418.774036	2351.679476	11087.223740	4418.774036	2351.679476
5	CentroidRMSE	0.054197	0.054454	0.061929	0.054197	0.054454	0.061929

AutoEncoder - > K_mean Test

```
from keras.models import Model
from keras.layers import Input, Dense, Dropout ,BatchNormalization

def create_simple_AE(X,Y, enc_size, activation_H, activation_out):
    in_layer = Input(shape=(X.shape[1],))
    enc_layer = Dense(enc_size, activation=activation_H)(in_layer)
    dec_layer = Dense(X.shape[1], activation=activation_out)(enc_layer)

AE = Model(in_layer, dec_layer)
Enc = Model(in_layer, enc_layer)

AE.compile(optimizer='adam', loss='mean_squared_error')

x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.2, random_state=0)
AE.fit(x_train, x_train,
```

```
epochs=50,
         batch size=50,
         shuffle=True,
        verbose=0.
        validation data=(x test, x test))
 return AE, Enc
def AE Kmean Test(data,encSize,activation,clusters num):
 # full Data
 X Original, Y Original = GetFeature Output(data)
 AE, Enc = create simple AE(X Original, Y Original, encSize, activation, activation)
 X = pd.DataFrame(Enc.predict(X Original))
 Clusterer = KMeans(n clusters=clusters num, random state=0).fit(X)
 Y Result = Clusterer.predict(X)
 Test Result = Evalution(X Original, Y Original.values, Y Result)
 # Train Test
 x train, x test, y train, y test = train test split(X Original, Y Original, test size=0.4, random state=0)
 AE, Enc = create simple AE(x train, y train, encSize, activation, activation)
 X train encoded = pd.DataFrame(Enc.predict(x train))
 X test encoded = pd.DataFrame(Enc.predict(x test))
 Clusterer = KMeans(n clusters=clusters num, random state=0).fit(X train encoded)
 Y Result = Clusterer.predict(X test encoded)
 result train test = Evalution(X Original.iloc[x test.index], y test, Y Result)
 # K-Folds
 KfordScorer = []
 k fold = StratifiedKFold(n splits=5)
 for train indices, test indices in k fold.split(X Original, Y Original):
   AE, Enc = create simple AE(X Original.iloc[train indices], Y Original.iloc[train indices], encSize, activation, activation)
   X train encoded = pd.DataFrame(Enc.predict(X Original.iloc[train indices]))
   X test encoded = pd.DataFrame(Enc.predict(X Original.iloc[test indices]))
   Clusterer = KMeans(n clusters=clusters num, random state=0).fit(X train encoded)
   Y Result = Clusterer.predict(X test encoded)
   y test = Y Original.iloc[test indices]
   resultTemp = Evalution(X Original.iloc[test indices], y test, Y Result)
   KfordScorer.append(resultTemp)
 results KFolds= np.mean(KfordScorer,axis=0)
```

```
return Test Result, result train test, results KFolds
```

```
r1,r2,r3 = AE_Kmean_Test(DataFinal,50,'sigmoid',2)
myResultsDF['AE_KMean_Full'] = r1
myResultsDF['AE_KMean__Train_Test'] = r2
myResultsDF['AE_KMean__K-folds'] = r3
myResultsDF
```



	Metric	Kmean_Full	Kmean_FTrain_Test	Kmean_FK- folds	PCA_Full	PCA_Train_Test	PCA_K- folds	AE_KMean_Fu
0	completeness_score	0.082771	0.083151	0.084843	0.082771	0.083151	0.084843	0.0000
1	adjusted_rand_score	0.199819	0.200140	0.201305	0.199819	0.200140	0.201305	0.0007
2	fowlkes_mallows_score	0.723365	0.723724	0.723783	0.723365	0.723724	0.723783	0.5751
3	silhouette_score	0.109456	0.109493	0.115697	0.109456	0.109493	0.115697	8800.0
4	calinski_harabaz_score	11087.223740	4418.774036	2351.679476	11087.223740	4418.774036	2351.679476	744.8719
5	CentroidRMSE	0.054197	0.054454	0.061929	0.054197	0.054454	0.061929	0.0348

AutoEncoder with SoftMax layer

```
def create_softmax_AE(X,Y, enc_size, activation_H, activation_out):
    in_layer = Input(shape=(X.shape[1],))
    enc_layer = Dense(enc_size, activation=activation_H)(in_layer)
    enc_layer = BatchNormalization()(enc_layer)
    enc_layer= Dropout(0.7)(enc_layer)
    enc_layer = Dense(Y.value_counts().size, activation='softmax')(enc_layer)
    enc_layer = BatchNormalization()(enc_layer)
    enc_layer = Dropout(0.7)(enc_layer)
    enc_layer = Dropout(0.7)(enc_layer)
    dec_layer = Dense(enc_size, activation=activation_out)(enc_layer)
```

```
dec layer = BatchNormalization()(dec layer)
 dec layer = Dropout(0.7)(dec layer)
 dec layer = Dense(X.shape[1], activation=activation out)(dec layer)
 AE = Model(in layer, dec layer)
 Enc = Model(in layer, enc layer)
 AE.compile(optimizer='adam', loss='mean squared error')
 x train, x test, y train, y test = train test split(X, Y, test size=0.2, random state=0)
 AE.fit(x train, x train,
        epochs=50,
        batch size=50,
         shuffle=True.
        verbose=0,
        validation data=(x test, x test))
 return AE, Enc
def AE SoftMax Test(data,encSize,activation):
 # full Data
 X Original, Y Original = GetFeature Output(data)
 AE, Enc = create softmax AE(X Original, Y Original, encSize, activation, activation)
 EncOutput = pd.DataFrame(Enc.predict(X Original))
 Y Result = EncOutput.idxmax(axis=1)
 Test Result = Evalution(X Original, Y Original.values, Y Result)
 # Train Test
 x train, x test, y train, y test = train test split(X Original, Y Original, test size=0.4, random state=12)
 AE, Enc = create softmax AE(x train, y train, encSize, activation, activation)
 EncOutput = pd.DataFrame(Enc.predict(x test))
 Y Result = EncOutput.idxmax(axis=1)
 result train test = Evalution(X Original.iloc[x test.index], y test, Y Result.values)
 # K-Folds
 KfordScorer = []
 k fold = StratifiedKFold(n splits=5)
 for train indices, test indices in k fold.split(X Original, Y Original):
   AE, Enc = create softmax AE(X Original.iloc[train indices], Y Original.iloc[train indices], encSize, activation, activation)
   EncOutput = pd.DataFrame(Enc.predict(X Original.iloc[test indices]))
   Y Result = EncOutput.idxmax(axis=1)
   #print("C2" ,Y Result.value counts())
   y test = Y Original.iloc[test indices]
```

```
#print("C1",y_test.value_counts())
    resultTemp = Evalution(X_Original.iloc[test_indices], y_test, Y_Result.values)
    KfordScorer.append(resultTemp)
    results_KFolds= np.mean(KfordScorer,axis=0)

    return Test_Result,result_train_test,results_KFolds

r1,r2,r3 = AE_SoftMax_Test(DataFinal,50,'sigmoid')

myResultsDF['AE_Softmax_Full'] = r1
    myResultsDF['AE_Softmax_Train_Test'] = r2
    myResultsDF['AE_Softmax_K-folds'] = r3
    myResultsDF
```

8

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3445: calling drough the drough tensorflow in the drough tensor of the drough

Please use `rate` instead of `keep prob`. Rate should be set to `rate = 1 - keep prob`.

	Metric	Kmean_Full	Kmean_FTrain_Test	Kmean_FK- folds	PCA_Full	PCA_Train_Test	PCA_K- folds	AE_KMean_Fu
0	completeness_score	0.082771	0.083151	0.084843	0.082771	0.083151	0.084843	0.0000
1	adjusted_rand_score	0.199819	0.200140	0.201305	0.199819	0.200140	0.201305	0.0007
2	fowlkes_mallows_score	0.723365	0.723724	0.723783	0.723365	0.723724	0.723783	0.5751
3	silhouette_score	0.109456	0.109493	0.115697	0.109456	0.109493	0.115697	8800.0
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5	CentroidRMSE	0.054197	0.054454	0.061929	0.054197	0.054454	0.061929	0.0348

Saving Results

```
myResultsDF.to_csv("Rain_evaluation_results.csv")
```