```
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('winequality-red.csv')
print('The Shape of The Data ',mData.shape)
mData.info()
    The Shape of The Data (1599, 12)
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 1599 entries, 0 to 1598
    Data columns (total 12 columns):
    fixed acidity
                             1599 non-null float64
    volatile acidity
                             1599 non-null float64
    citric acid
                             1599 non-null float64
    residual sugar
                             1599 non-null float64
    chlorides
                             1599 non-null float64
    free sulfur dioxide
                             1599 non-null float64
    total sulfur dioxide
                             1599 non-null float64
    density
                             1599 non-null float64
    рΗ
                             1599 non-null float64
    sulphates
                             1599 non-null float64
    alcohol
                             1599 non-null float64
    quality
                             1599 non-null int64
    dtypes: float64(11), int64(1)
    memory usage: 150.0 KB
```

Preprocessing the Data

```
from sklearn import preprocessing
from sklearn.utils import resample
def Preprocessing Data(data):
 data['quality'] = data["quality"] - 3
 Y = data['quality']
 X = data.drop(columns=['quality'])
 scaler = preprocessing.MinMaxScaler().fit(X)
 x normalized = scaler.transform(X)
 newData = pd.DataFrame(columns= X.columns,data=x normalized)
 newData['quality'] = Y
 # -----Ballancing the Data set-----
 Categories = pd.Categorical(Y.astvpe('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],
                     replace=True,
                                       # sample with replacement
                     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 BallancedData , scaler
DataFinal, scaler = Preprocessing Data(mData)
→ total shape after up sampling = (4086, 12)
    (4086, 12)
```

Data Helping Functions

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

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

Results Variables

Evaluation Functions

```
#Centroid Evaluation Function
def CentroidRMSE(Features,Clutered1,Clustered2):
    #print("C1 IN ",Clutered1.shape)
    #print("C2 IN ",Clustered2.shape)

X1 = Features.copy()
    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)
```

С→

```
# 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, 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,6)
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	0.220732	0.196039	0.254787
	1 adjusted_rand_score		0.142946	0.112454	0.164138
	2	fowlkes_mallows_score	0.300667	0.276166	0.323162
	3	silhouette_score	0.234202	0.227988	0.239240
	4	calinski_harabaz_score	1032.076059	401.171639	206.378573
	5	CentroidRMSE	0.129906	0.144322	0.163664

PCA -> K_Mean Test

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,6,5)
myResultsDF['PCA Full'] = r1
myResultsDF['PCA Train Test'] = r2
```

С⇒

```
myResultsDF['PCA_K-folds'] = r3
myResultsDF
```

→		Metric	Kmean_Full	Kmean_FTrain_Test	Kmean_FK-folds	PCA_Full	PCA_Train_Test	PCA_K-folds
	0	completeness_score	0.220732	0.196039	0.254787	0.218227	0.214853	0.232714
	1	adjusted_rand_score	0.142946	0.112454	0.164138	0.141182	0.136143	0.140458
	2	fowlkes_mallows_score	0.300667	0.276166	0.323162	0.297863	0.293402	0.298038
	3	silhouette_score	0.234202	0.227988	0.239240	0.233839	0.227327	0.228807
	4	calinski_harabaz_score	1032.076059	401.171639	206.378573	1032.396724	401.099790	204.639195
	5	CentroidRMSE	0.129906	0.144322	0.163664	0.144000	0.125932	0.132044

AutoEncoder - > K_mean Test

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

Description backend.

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,10,'sigmoid',6)

myResultsDF['AE_KMean_Full'] = r1
myResultsDF['AE_KMean_Train_Test'] = r2
myResultsDF['AE_KMean_K-folds'] = r3
myResultsDF
```

python3.6/dist-packages/tensorflow/python/framework/op_def_library.py:263: colocate_with (from tensorflow.python.:

placer.

>/python3.6/dist-packages/tensorflow/python/ops/math_ops.py:3066: to_int32 (from tensorflow.python.ops.math_ops) is

AE_KMeanK- folds	AE_KMeanTrain_Test	AE_KMean_Full	PCA_K- folds	PCA_Train_Test	PCA_Full	Kmean_FK- folds	an_FTrain_Test
0.260030	0.198721	0.231459	0.232714	0.214853	0.218227	0.254787	0.196039
0.170808	0.130478	0.166723	0.140458	0.136143	0.141182	0.164138	0.112454
0.319184	0.284217	0.309826	0.298038	0.293402	0.297863	0.323162	0.276166
0.202508	0.171895	0.194350	0.228807	0.227327	0.233839	0.239240	0.227988
195.884810	357.920717	994.602390	204.639195	401.099790	1032.396724	206.378573	401.171639
0.124141	0.114560	0.132300	0.132044	0.125932	0.144000	0.163664	0.144322

AutoEncoder with SoftMax layer

```
from keras.activations import softmax

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)
  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=0)
  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, 10, 'sigmoid')
myResultsDF['AE Softmax Full'] = r1
myResultsDF['AE Softmax Train Test'] = r2
myResultsDF['AE Softmax K-folds'] = r3
myResultsDF
```

□→ WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3445: calling d Instructions for updating:

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_Full
0	completeness_score	0.220732	0.196039	0.254787	0.218227	0.214853	0.232714	0.231459
1	adjusted_rand_score	0.142946	0.112454	0.164138	0.141182	0.136143	0.140458	0.166723
2	fowlkes_mallows_score	0.300667	0.276166	0.323162	0.297863	0.293402	0.298038	0.309826
3	silhouette_score	0.234202	0.227988	0.239240	0.233839	0.227327	0.228807	0.19435(
4	calinski_harabaz_score	1032.076059	401.171639	206.378573	1032.396724	401.099790	204.639195	994.602390
5	CentroidRMSE	0.129906	0.144322	0.163664	0.144000	0.125932	0.132044	0.13230(

Saving Results

myResultsDF.to_csv("evaluation_results_UpSampling.csv")