baisravanhc

Assignment 4

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POSTED BY BAISRAVAN IN UNCATEGORIZED

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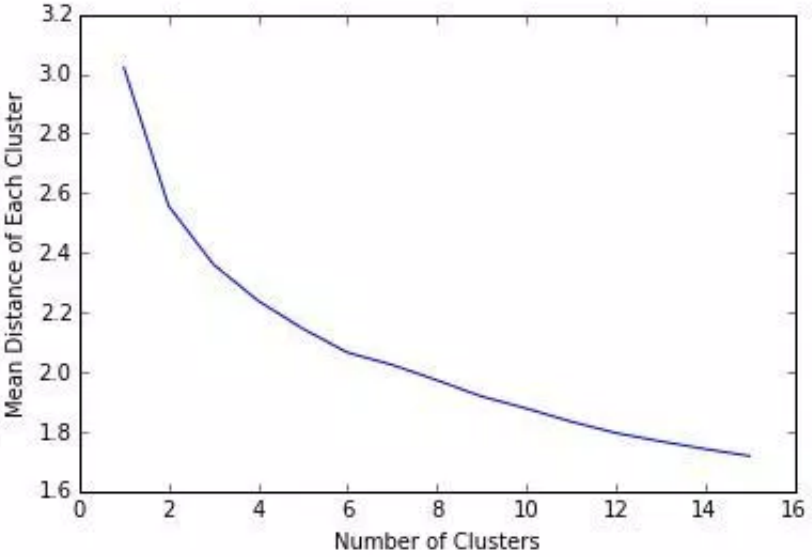
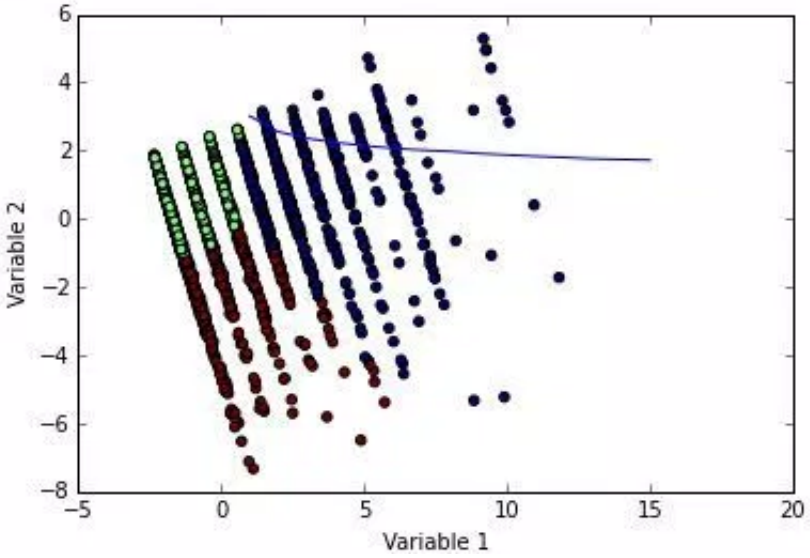
In this assignment, I am clustering data with clustering variables: marijuana use, alcohol problem, deviant behavior scale, violent behavior scale, depression scale, self esteem scale, school connectedness scale and parental presence scale.

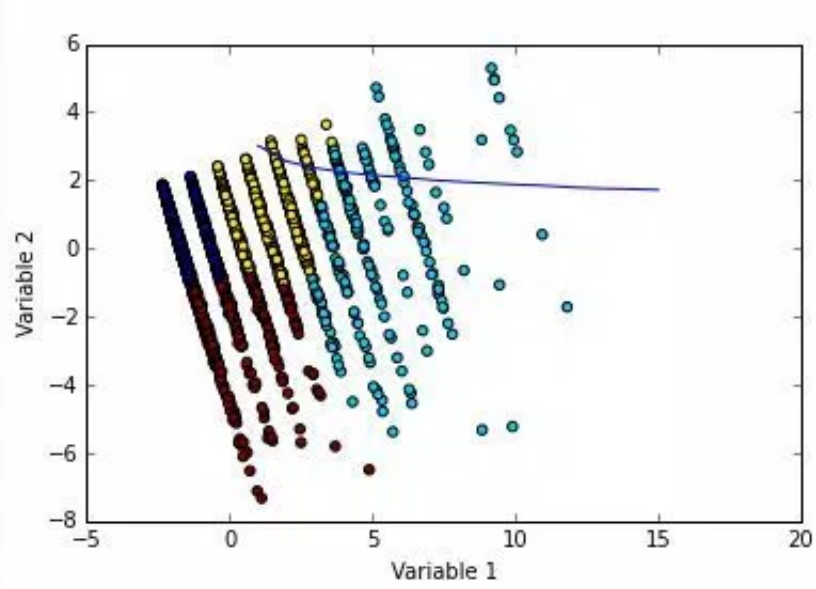
First, the data was processed and the graph showing the mean distance of the clusters with the number of clusters (elbow plot) was plotted. The plot is shown below:



Then I chose 3 clusters and the result is shown below:



It seemed like another cluster would be beneficial:



Clearly 4 clusters seemed to classify the data better (It was seen that 5 clusters was not doing a good jobs).

The 4 clusters looks like this:

cluster	index	MAREVER1	ALCPROBS1	DEVIANT1	VIOL1	DEP1
0	3346.292984	-0.386755	-0.300552	-0.406206	-0.301898	-0.329810
1	3449.274590	0.698189	0.636246	0.723714	0.534342	0.532673
2	3231.878702	-0.242238	-0.258097	-0.215692	-0.182535	-0.181774
3	3223.692929	1.029389	0.745582	0.944463	0.740005	0.718614

cluster	ESTEEM1	SCHCONN1	PARPRES
0	0.234861	0.338298	14.797224
1	-0.337658	-0.560404	8.688525
2	0.134452	0.210997	11.988717
3	-0.494573	-0.820696	14.185859

It can be seen that 0 and cluster 2 were more or less similar in nature while cluster 1 and 3 showed higher likelihood of alcohol problem, violent and deviant behavior, depression, etc. These were low for the clusters 0 and 2.

CODE:

```
Spyder Editor

This is a temporary script file.
"""

import numpy as np
from pandas import DataFrame
import pandas as pd
import matplotlib.pylab as plt
from sklearn.cross_validation import train_test_split
from sklearn import preprocessing
from sklearn.cluster import KMeans
from scipy.spatial.distance import cdist
from sklearn.decomposition import PCA
data = pd.read_csv('tree_addhealth.csv')

data.columns = map(str.upper,data.columns)

data = data.dropna()

cluster_var = data[['MAREVER1','ALCPROBS1','DEVIANT1','VIOL1',
'DEP1','ESTEEM1','SCHCONN1','PARPRES']]

cluster_var.columns = map(str.upper,cluster_var.columns)
cluster_var.describe()
clusters = cluster_var.copy()
for i in range(0,len(cluster_var.columns)-1):
clusters[cluster_var.columns[i]] =
preprocessing.scale(clusters[cluster_var.columns[i]].astype('float64'))

cluster_train,cluster_test = train_test_split(clusters,test_size = 0.4,random_state = 123)

cluster_range = range(1,16)
mean_dist = []

for k in cluster_range:
model= KMeans(n_clusters = k)
model.fit(cluster_train)
mean_dist.append(sum(np.min(cdist(cluster_train, model.cluster_centers_, 'euclidean'),
axis=1))
/ cluster_train.shape[0])

plt.plot(cluster_range,mean_dist)
plt.xlabel('Number of Clusters')
plt.ylabel('Mean Distance of Each Cluster')
```

```
model4 = KMeans(n_clusters = 3)
model4.fit(cluster_train)

pca2 = PCA(2)
plot_column = pca2.fit_transform(cluster_train)
plt.scatter(x = plot_column[:,0],y=plot_column[:,1], c=model4.labels_,)
plt.xlabel('Variable 1')
plt.ylabel('Variable 2')

cluster_train.reset_index(level=0, inplace=True)
cluster_list=list(cluster_train['index'])

labels=list(model4.labels_)

newlist=dict(zip(cluster_list, labels))
newclus=DataFrame.from_dict(newlist, orient='index')
newclus.columns = ['cluster']
newclus.reset_index(level=0, inplace=True)
merged_train=pd.merge(cluster_train, newclus, on='index')
merged_train.head(n=100)
merged_train.cluster.value_counts()

clustergrp = merged_train.groupby('cluster').mean()
```