

Video Link: <https://www.loom.com/share/618271486570498289ab449e8f89dcba>

1. Apply K means clustering in this data set provided below:
<https://umkc.box.com/s/s15r7m0gnxu7b1s2kaobvc5w7da2nc1c>
 - a. Remove any null values by the mean.
 - b. Use the elbow method to find a good number of clusters with the KMeans algorithm
2. Calculate the silhouette score for the above clustering
3. Try feature scaling to see if it will improve the Silhouette score
4. Apply PCA on the same dataset.
5. *** Bonuspoints Apply kmeans algorithm on the PCA result and report your observation if the score improved or not?

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ICP6 > kmeansclustering_123_Bonus.py
kmeansclustering_123_Bonus.py x Clusterusing.py x
1  # Importing Libraries
2  import pandas as pd
3  import matplotlib.pyplot as plt
4  import warnings
5  warnings.filterwarnings("ignore")
6  from sklearn.cluster import KMeans
7  from sklearn import metrics
8  from sklearn.preprocessing import StandardScaler
9  from sklearn.decomposition import PCA
10
11  # Reading the data & identifying the feature to form the clusters
12  customer = pd.read_csv('CC.csv')
13  x = customer.iloc[:, 1:17]
14  y = customer.iloc[:, -1] #Last column of data frame
15
16  #1a Computing mean of data containing null values to replace them with its mean
17  MeanNA = customer.loc[:, "MINIMUM_PAYMENTS"].mean()
18  print('Mean of Minimum Payments is ', MeanNA)
19  x = x.fillna(MeanNA)
20
21  #1b Elbow point computation to determine good number of clusters
22  wcss = []
23  for i in range(1, 11):
24      kmeans = KMeans(n_clusters=i, max_iter=300, random_state=0)
25      kmeans.fit(x)
26      wcss.append(kmeans.inertia_)
27
28  #Plotting the elbow point on graph
29  plt.plot(range(1, 11), wcss)
30  plt.title('Elbow Method')
31  plt.xlabel('Number of Clusters')
32  plt.ylabel('Wcss')
```

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# Performing K-Means clustering on the data available
nclusters = 4 #This is the K in mean
km = KMeans(n_clusters=nclusters)
km.fit(x)

#2 Evaluation of the clusters and silhouette score
y_cluster_KMeans = km.predict(x)
score = metrics.silhouette_score(x, y_cluster_KMeans)
print('Silhouette Score of the Clusters is ', score)

#3 Feature Scaling

scaler = StandardScaler()# Fit on training set only.
scaler.fit(x)

# Apply transform to both the training set and the test set.
x= scaler.transform(x)
X_scaled_array=scaler.transform(x)
X_scaled=pd.DataFrame(X_scaled_array)
x=X_scaled

##building the model
nclusters = 4 # this is the k in kmeans
km = KMeans(n_clusters=nclusters)
km.fit(x)

# predict the cluster for each data point
y_cluster_kmeans = km.predict(x)
score = metrics.silhouette_score(x, y_cluster_kmeans)
print('Silhouette Score of the Clusters after Scaling is ', score)

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# Standardization of the data
scaler = StandardScaler()
scaler.fit(x)


# Projecting data on reduced dimension
x_scaler = scaler.transform(x)

# Performing Principle Component Analysis (PCA)
pca = PCA(2)
x_pca = pca.fit_transform(x_scaler)
df2 = pd.DataFrame(data=x_pca) #printdf2

# Bonus: KMeans on PCA
# Performing K-Means clustering on the PCA data
nclusters = 4
km = KMeans(n_clusters=nclusters)
km.fit(x_pca)

# Evaluation of the clusters accuracy
y_cluster_KMeans = km.predict(x_pca)
score = metrics.silhouette_score(x_pca, y_cluster_KMeans)
print('Silhouette Score of the Clusters after applying Kmean on PCA is ', score)

```

kmeansclustering_123_Bonus (1) 

```

C:\ProgramData\Anaconda3\python.exe "C:/Users/1d630534/Documents/Anurag Projects/Spring 2020/Python with DL/ICP6/kmeansclustering_123_Bonus.py"
Mean of Minimum Payments is 864.2065423050814
Silhouette Score of the Clusters is 0.4656739200759652
Silhouette Score of the Clusters after Scaling is 0.2962453046974821
Silhouette Score of the Clusters after applying Kmean on PCA is 0.4096478264292878

Process finished with exit code 0

```