Assignment

Question 1: Assignment Summary

Briefly describe the "Clustering of Countries" assignment that you just completed within 200-300 words. Mention the problem statement and the solution methodology that you followed to arrive at the final list of countries. Explain your main choices briefly (what EDA you performed, which type of Clustering produced a better result and so on)

Answer:

Our main objective for this assignment is to find the countries that are in dire need of aid. Our job is to find those countries using socio-economic and heath factors which will show overall development of the country. So, in order to do so first I analyze the dataset – It has total 167 countries with no missing values, I transformed some of the variables like health, imports and exports from % of GDP to absolute values as it makes more sense to the dataset. Then I visualized the data by finding the correlation between variables to check the multicollinearity and found that most of the variables are having multicollinearity. Now I checked the outliers of the dataset and I found there are outliers on all the variables. It was found that all the features consist of outliers. So I used soft capping method to remove outliers i.e. removed data points above 0.99 quantile. I checked this by describing the dataset in percentiles and by plotting boxplot. A Hopkins statistic was applied on the data to check if the data can be divided into clusters. It gave a value of 0.89 which is pretty good for clustering. So I went ahead with data modelling.

Then, before feeding the data into the model, I used standard scaler to scale down the variable which makes it easy for the model to interpret. Then, K-Means algorithm was used to cluster the data. Elbow curve/SSD and silhouette analysis was done on the data to find the optimal number of k. The optimal value was found out to be k=3. So the data was divided into 3 clusters by using k-means clustering. Then second approach was the hierarchical clustering in which single and complete linkage method was used to plot the dendrogram of the data. After this, complete linkage method proved to be appropriate for clustering. So a threshold value of 10 and then 8 was considered and the clusters were formed accordingly. After analyzing the hierarchical clusters, a conclusion was made that the data needs to be segmented in 3 clusters

After performing the clustering, the countries which were in need of financial aid were:

- Burundi
- Liberia
- Congo, Dem. Rep.

- Niger
- Sierra Leone

Question 2: Clustering

- a) Compare and contrast K-means Clustering and Hierarchical Clustering.
 - b) Briefly explain the steps of the K-means clustering algorithm.
- c) How is the value of 'k' chosen in K-means clustering? Explain both the statistical as well as the business aspect of it.
 - d) Explain the necessity for scaling/standardisation before performing Clustering.
 - e) Explain the different linkages used in Hierarchical Clustering.

a)

| K means clustering | Hierarchical Clustering |
|--|---|
| We need to have desired number of clusters | We can decide the number of clustersafter |
| ahead of time. | completion of plotting dendrogram by cutting |
| | the dendrogram at different heights |
| It is a collection of data points in one cluster | Clusters have tree like structures and most |
| which are similar between them and not | similar clusters are first combine which |
| similar data points belongs to another cluster | continues until we reach a single branch. |
| Works very good in large dataset | Works well in small dataset and not good |
| | with large dataset |
| The main drawback of k-Means is it doesn't | Outliers are properly explained in hierarchical |
| evaluate properly outliers. | clustering |
| K-means only used for numerical. | Hierarchical clustering is used when we have |
| | variety of data as it doesn't require to |
| | calculate any distance. |

b)

- Step 1: Randomly select K points as initial centroids.
- Step 2: All the data points closet to the centroid will create cluster center according to Euclidean distance function.
- Step 3: Once we assign all the points to each of k clusters, we need to update the cluster centers or centroid of that cluster created.
- Step 4: Repeat 2,3 steps until cluster centers reach convergence.

'K' value is chosen randomly in K-Means clustering based on statistical aspect. From business aspect, we need to first understand the dataset and based on that we decide number of 'k'. for example, we have a dataset of variables like 'pen', 'pencil', 'books', 'notebooks', 'mobiles', 'charger', 'laptop'. Now if we want to have k values based on statistical aspect, we can use silhouette score to determine that but based on business aspect, after viewing the dataset we can easily make cluster = 2, one in electronics category and another non-electronics.

D)

It is definitely a good idea to do scaling/standardisation because our variables may have units at different scale and as our method stresses more on calculation of direction of space or distance, so if we have one variable with high scale units then while calculating for k-Means or hierarchical it will create a big difference as the clusters will tend to move with the variables having greater values or variances. By applying standardisation/scaling will increase the performance of our model.

e)

Linkage is a technique used in Agglomerative Clustering. Linkage helps us to merge two data points into one using below linkage technique. Single linkage: The distance between two clusters is calculated by the minimum distance between two points from each cluster. Complete linkage: The distance between two clusters is calculated by the maximum distance between two points from each cluster. Average linkage: The distance between two clusters is the average distance between every point of one cluster to the another every point of other cluster. Ward linkage: The distance between clusters is calculated by the sum of squared differences with all clusters.