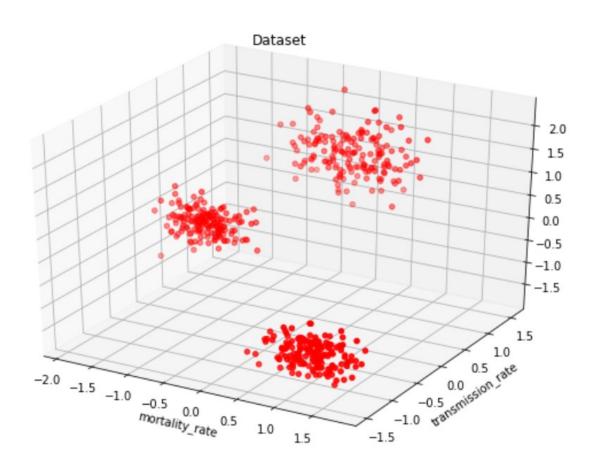
PROJECT-3 DC1

Coronavirus Data Clustering using Complete Linkage Hierarchical Clustering Technique



Scatter plot of the given Dataset

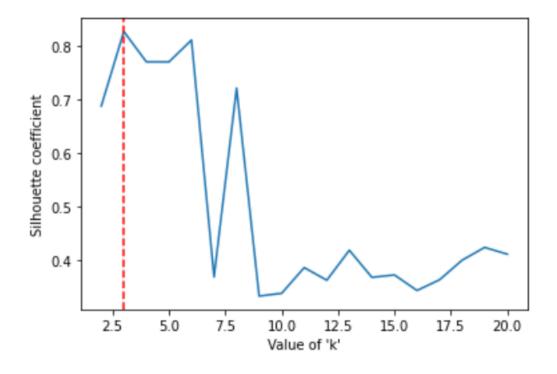
1

Optimal Number of Clusters

The optimal number of clusters obtained after calculating the silhouette coefficient for k = 3, 4, 5, and 6 is **3.** The values obtained are listed below.

```
Silhouette Coefficient for k=3 is 0.8700689668748269
Silhouette Coefficient for k=4 is 0.8629278588557886
Silhouette Coefficient for k=5 is 0.8658016516543212
Silhouette Coefficient for k=6 is 0.24924153711362315
Optimal number of clusters =3
```

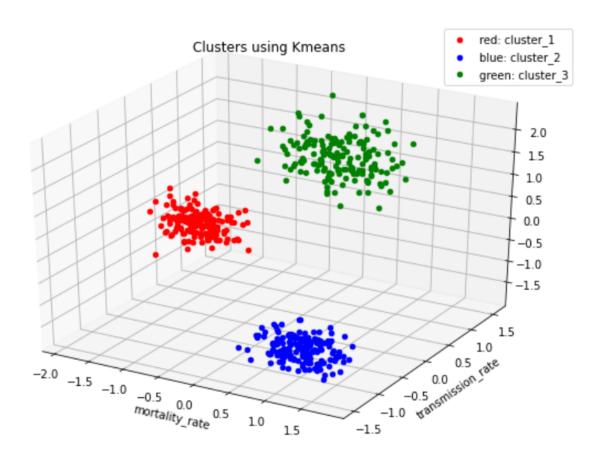
The graph below records silhouette coefficients for various values of k. The dashed line is of k = 3, which has the maximum value of silhouette coefficient.



Graph of Silhouette coefficient vs Value of 'k'

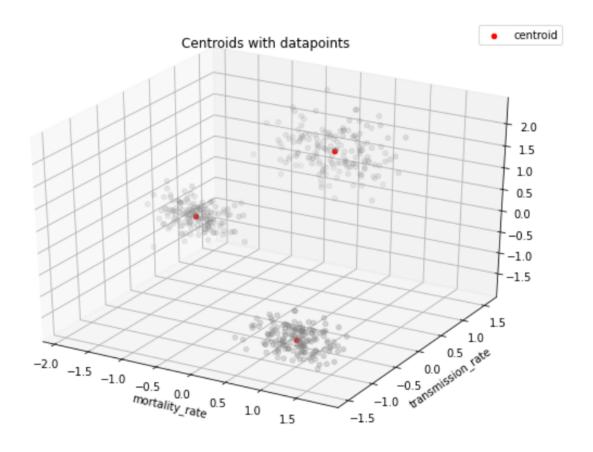
Analysis of Silhouette Coefficient (in step - 2)

The silhouette coefficient is a metric used to calculate the goodness of a clustering technique. Its value ranges from -1 to 1. The values closer to one are considered good. In my case, I obtained a silhouette coefficient of 0.8700689668748269 for k = 3 using k-means Clustering. The value is close to one and hence is good. We can also visualize the clusters and centroids for the algorithm.



Scatter plot of the clusters categorized by K-means

We can see that my algorithm has categorized all three clusters perfectly. Hence we got a good value of silhouette coefficient. But the value is not exactly 1 because the silhouette coefficient depends upon both **cohesion** and **separation** of clusters and we can see that the clusters are somewhat spread out hence leading to a decrement in cohesion value and therefore the whole coefficient.



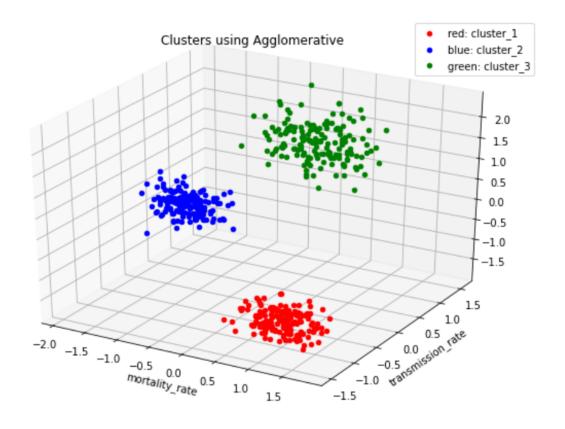
Scatter plot of datapoints with centroids marked

Analysis of Jaccard Similarity Scores (in step - 4)

The Jaccard Similarity Scores obtained for all the three mappings from k-means deduced clusters to agglomerative hierarchical clusters is **1**. It is because the Jaccard similarity score is the ratio of intersection of clusters and union of clusters, and the clusters that I received from both algorithms are the same. So union(A,B) = intersection(A,B), which implies their ratio will be one.

```
Cluster 0 of kmeans is mapped to Cluster 1 of agglomerative Cluster 1 of kmeans is mapped to Cluster 0 of agglomerative Cluster 2 of kmeans is mapped to Cluster 2 of agglomerative Jaccard Similarity Score for the 0 -> 1 mapping: 1.0 Jaccard Similarity Score for the 1 -> 0 mapping: 1.0 Jaccard Similarity Score for the 2 -> 2 mapping: 1.0
```

We can also visualize that the clusters obtained from agglomerative are the same as that of k-means.



Scatter plot of the clusters categorized by Agglomerative clustering

THE END