Below are the codes after the Silhouettee analysis (end of class on November 4). It will be most helpful if you enter these codes in your colab notebook before our class on Nov 11, so that we can spend the class hours on explaining the concepts and how the codes work.

The codes below are copied from https://scikit-

<u>learn.org/stable/auto_examples/cluster/plot_agglomerative_dendrogram.html#sphx-glr-auto-examples-cluster-plot-agglomerative-dendrogram-py</u>

```
[34] from scipy.cluster.hierarchy import dendrogram

def plot_dendrogram(model, **kwargs):
    # Create linkage matrix and then plot the dendrogram

# create the counts of samples under each node
    counts = np.zeros(model.children_.shape[0])
    n_samples = len(model.labels_)
    for i, merge in enumerate(model.children_):
        current_count = 0
        for child_idx in merge:
            if child_idx < n_samples:
                current_count += 1 # leaf node
        else:
                current_count += counts[child_idx - n_samples]
        counts[i] = current_count

linkage_matrix = np.column_stack(
        [model.children_, model.distances_, counts]
).astype(float)

# Plot the corresponding dendrogram
    dendrogram(linkage_matrix, **kwargs)</pre>
```

```
[35] agg_cluster = AgglomerativeClustering(distance_threshold=0, n_clusters= None).fit(Xs_100F)
    plt.title('Hierarchical cluster Dendrogram')
    plot_dendrogram(agg_cluster, truncate_mode='level', p=10)
    plt.xlabel('Number of points in node (or index of point if no parenthesis')
    plt.show()

[36] agg_7 = AgglomerativeClustering(n_clusters=7).fit(Xs_10DF)
    gdffract2020['agg_7'] = agg_7.labels_
    gdffract2020['agg_7'] = agg_7.labels_
    gdffract2020.plot(column='agg_7', categorical=True, cmap='Spectral', legend=True, legend_kwds={'loc': 'lower right'})

[37] silhouette_score(Xs_100F, agg_7.labels_)

[38] aff = AffinityPropagation(preference=-500, random_state=10).fit(Xs_10DF)
    n_clusters = len(aff.cluster_centers_indices_)
    silhouette = silhouette_score(Xs_10DF, aff.labels_)
    print('estimated number of clusters: %d' % n_clusters)
    print('estimated number of clusters: %d' % n_clusters)
    print('Silhouettee coefficient: %0.3f' % silhouette)

[39] aff.cluster_centers_indices_
```

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[40] idxlist = aff.cluster_centers_indices_.tolist()
       idxlist
 [41] prototype = gdfTract2020.iloc[idxlist]
       prototype
[42] fig, ax = plt.subplots(figsize=(10,8))
      gdfTract2020['aff_2'] = aff.labels_
gdfTract2020.plot(column='aff_2', ax=ax)
prototype.plot(color='red', ax=ax)
 [44] gdfTract2020.columns
 [45] agg_2 = AgglomerativeClustering(n_clusters=2).fit(Xs_10DF)
       gdfTract2020['agg_2'] = agg_2.labels_
gdfTract2020.plot(column='agg_2')
[46] !pip install pysal
!pip install shapely
[47] from pysal.lib import weights
[48] wrook = weights.contiguity.Rook.from_dataframe(gdfTract2020)
      wqueen = weights.contiguity.Queen.from_dataframe(gdfTract2020)
[49] wrook.neighbors
[50] pd.DataFrame(*wrook.full()).astype(int)
[51] wrook.nonzero
[52] agg_spatial = AgglomerativeClustering(n_clusters=5, connectivity=wqueen.sparse).fit(Xs_10DF)
    gdfTract2020['agg_spatial'] = agg_spatial.labels_
    gdfTract2020.plot(column='agg_spatial')
      silhouette = silhouette_score(Xs_10DF, agg_spatial.labels_)
print('Silhouettee coefficient: %0.3f' % silhouette)
[54] !pip install sklearn_som
[55] from sklearn_som.som import SOM
      X = Xs_10DF.to_numpy()
      som_cls = SOM(m=1, n=2, dim=10).fit_predict(X)
[56] som_cls
[57] silhouette = silhouette_score(Xs_10DF, som_cls)
[58] gdfTract2020['som2'] = som_cls
```