day61-hierarchical-clustering

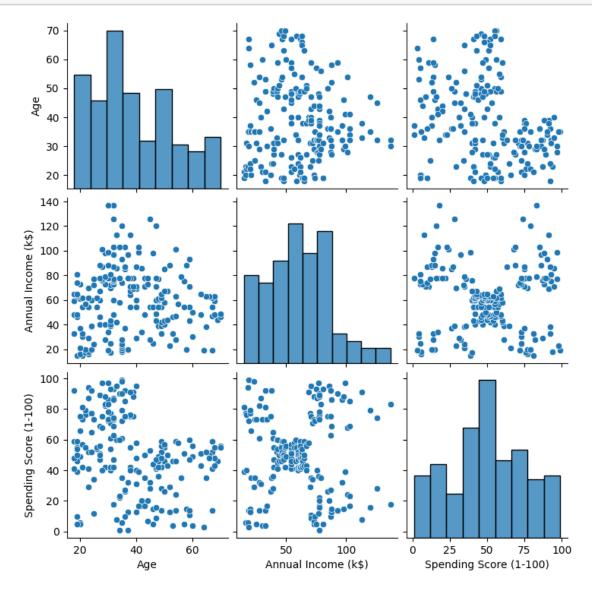
December 9, 2023

Day61 Hierarchical Clustering By: Loga Aswin

```
[8]: # import libraries
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      from sklearn.cluster import AgglomerativeClustering
      from sklearn import preprocessing
      from scipy.cluster.hierarchy import dendrogram, linkage
 [9]: # load dataset
      df = pd.read_csv('Mall_Customers.csv')
[10]: df.head()
[10]:
         CustomerID
                       Genre
                                   Annual Income (k$)
                                                        Spending Score (1-100)
                              Age
                  1
                       Male
                               19
                                                                             39
                                                    15
                  2
                       Male
                                                                             81
      1
                               21
                                                    15
      2
                  3 Female
                               20
                                                                              6
                                                    16
                  4 Female
                                                                             77
                               23
                                                    16
                     Female
                               31
                                                    17
                                                                             40
[11]: df.isnull().sum()
[11]: CustomerID
                                 0
      Genre
                                 0
                                 0
      Age
      Annual Income (k$)
                                 0
      Spending Score (1-100)
      dtype: int64
[12]: df.describe()
[12]:
             CustomerID
                                      Annual Income (k$)
                                                           Spending Score (1-100)
                                 Age
      count
             200.000000
                         200.000000
                                               200.000000
                                                                        200.000000
      mean
             100.500000
                           38.850000
                                                60.560000
                                                                         50.200000
      std
              57.879185
                           13.969007
                                               26.264721
                                                                         25.823522
      min
               1.000000
                           18.000000
                                                15.000000
                                                                          1.000000
```

```
25%
        50.750000
                    28.750000
                                         41.500000
                                                                  34.750000
50%
       100.500000
                    36.000000
                                         61.500000
                                                                  50.000000
       150.250000
                    49.000000
                                         78.000000
                                                                  73.000000
75%
       200.000000
                    70.000000
                                        137.000000
                                                                  99.000000
max
```

```
[13]: sns.pairplot(df[['Age', 'Annual Income (k$)', 'Spending Score (1-100)']]) plt.show()
```

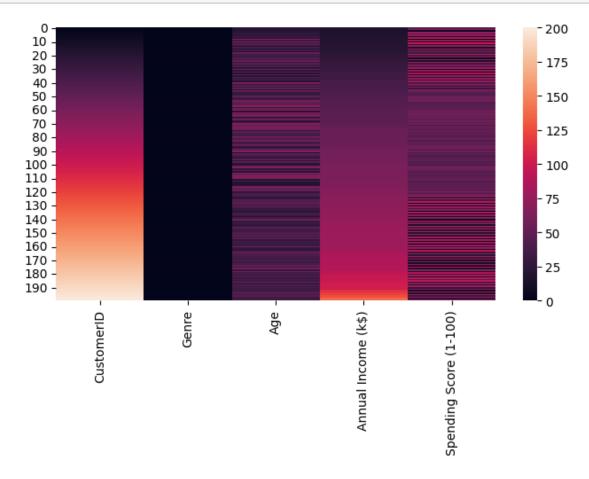


```
[14]: label_encoder = preprocessing.LabelEncoder()

df['Genre'] = label_encoder.fit_transform(df['Genre'])
    df.head()
```

```
[14]:
         CustomerID Genre
                                    Annual Income (k$)
                                                           Spending Score (1-100)
                               Age
      0
                                19
                           1
                    2
                           1
                                21
                                                      15
      1
                                                                                 81
      2
                    3
                           0
                                20
                                                      16
                                                                                  6
      3
                    4
                                23
                                                                                 77
                           0
                                                      16
                   5
      4
                           0
                                31
                                                      17
                                                                                 40
```

```
[15]: plt.figure(1, figsize = (8 ,4))
sns.heatmap(df)
plt.show()
```

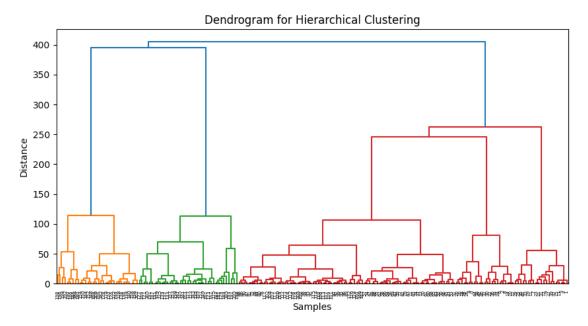


Feature Selection:

Hierarchical Clustering

using the 'ward' linkage method, creating the linkage matrix.

Dendrogram for Hierarchical Clustering



Agglomerative Clustering Model Training

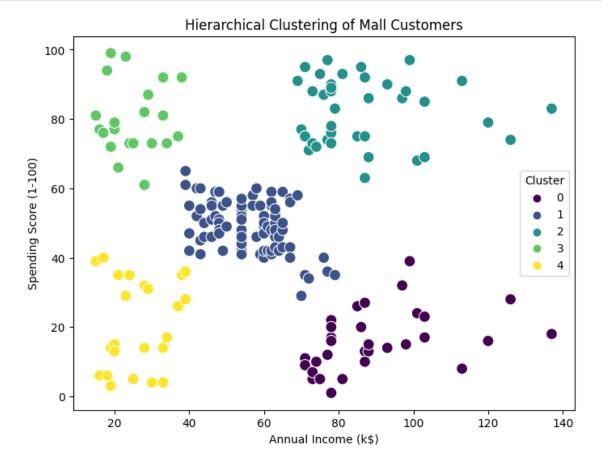
Training the model with 5 clusters using Euclidean distance and 'ward' linkage.

```
[26]: hc = AgglomerativeClustering(n_clusters=5, affinity='euclidean', linkage='ward')
y_pred = hc.fit_predict(X)
```

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_agglomerative.py:983: FutureWarning:

Attribute `affinity` was deprecated in version 1.2 and will be removed in 1.4. Use `metric` instead

```
[30]: df['Cluster'] = y_pred y_pred
```



```
[29]: from sklearn import metrics
```

```
silhouette = metrics.silhouette_score(X, y_pred)
print(f"Silhouette Score: {silhouette}")
```

Silhouette Score: 0.5529945955148897

Plotting in 3D using Plotly

```
[31]: import plotly.graph_objs as go
     import plotly.offline as py
     sample = go.Scatter3d(
         x=df['Annual Income (k$)'], y=df['Spending Score (1-100)'], z=df['Age'],
       marker=dict( color=df['Cluster'], size=10, line=dict(color=df['Cluster'],
       ⇒width=12), opacity=0.8
                                 )
     data_plotly = [sample]
     layout = go.Layout(title='Clusters using Agglomerative Clustering',
         scene=dict(
             xaxis=dict(title='Annual Income (k$)'),
             yaxis=dict(title='Spending Score (1-100)'),
             zaxis=dict(title='Age')
         )
     fig = go.Figure(data=data_plotly, layout=layout)
     py.iplot(fig)
```

Our plot visualizes how customers are grouped based on these attributes after clustering.

By using Agglomerative Clustering on our mall_customer information, we grouped them based on how much they earn, spend, and their age.

So, This customer segmentation helps us discover better ways to sell products/services and grow our business.