K-Medoids clustering on socio-economic dataset In [6]: import pandas as pd from sklearn extra.cluster import KMedoids from sklearn.metrics import silhouette score df = pd.read excel('Preprocessed data standardscaler.xlsx') # Dropping the ID Column, as it is not required in clustering. new df = df.drop(columns=['ID']) new df Out[6]: **Education Education** Income Helme Income Income **Normalised Gender** Income Experience Experience Experience Experience Experience Experience **Higher Income** 60-(<Class (Class Graduate 20-40-(Full <20k >80k (2-5yr) (5-10yr) (10-20yr) (20-30yr) (>30yr) Age (Yrs) (Male) Edu (<1yr) (1-2yr) 40k 60k 80k 10) 10-12) mask 0 0 0 0 0 -0.927352 0 0 0 1 0 0 0 0 0 1 0 0 0 0 0 0.387063 0 1 0 0 0 1 0 0 0 1 0 0 0 2 -0.739578 0 0 1 0 0 0 1 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 3 0.762611 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 -1.678446 4 -0.739578 0 0 0 0 0 0 0 1 0 0 476 0 0 1 0 0 -1.302899 0 0 0 0 0 0 0 0 1 0 0 0 0 477 0 0 1 0 0 0 0 0 0 0 0 0 478 2.077026 0 1 1 0 0 0 0 0 0 0 -0.551804 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 480 -0.551804 0 481 rows × 19 columns # Now the Silhouette Score method is used to determine the number of clusters to achieve the best possible clustering # Here the scores are calculated for $k = \{2,3,4,5,6\}$ scores = {} for i in range(2,7): kmedoids = KMedoids(n_clusters=i,method='pam',init='build',random_state=42) kmedoids.fit(new df) score = silhouette score(new df,kmedoids.labels_) scores[i]=score df scores = pd.DataFrame(list(scores.items()), columns=['no. of clusters', 'silhouette score']) df scores # since setting k = 5 yields the highest score, the number of clusters are chosen to be 5. # As can be seen in the table below: Out[7]: no. of clusters silhouette score 2 0 0.137599 3 0.152380 2 4 0.170523 5 3 0.172540 6 4 0.162440 # Now applying the K-Medoids algorithm. optimal_clusters = 5 kmedoids 1 = KMedoids(n clusters=optimal clusters, method='pam', init='build', random state=42) kmedoids_1.fit(new_df) list1 = list(kmedoids 1.labels) df clustered = new df.copy() df clustered['Cluster'] = list1 df clustered # As can be seen in the below dataset, a new 'Cluster' column has been added at the very end: **Education Education** Out[8]: Helme Income Income Income **Normalised Gender Higher Income** Income Experience Experience Experience Experience Experience Experience (<Class (Class Graduate 20-40-60-(Full Age (Yrs) (Male) Edu <20k >80k (2-5yr)(5-10yr) (10-20yr) (20-30yr) (>30yr) (<1yr) (1-2yr)10) 10-12) 40k 60k 80k mask 0 0 0 0 -0.927352 0 0 0 0 0 0 0 0 0 0 0 0 0 0.387063 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 2 -0.739578 0 0 1 0 0 1 0 0 1 0 3 0.762611 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 -1.678446 1 0 0 1 0 4 • • • • • • -0.739578 0 0 0 0 0 0 0 0 0 476 0 0 0 1 0 0 0 0 477 -1.302899 0 0 1 0 0 0 0 1 0 0 0 2.077026 0 0 1 0 0 0 0 1 0 0 0 0 0 0 1 0 478 0 0 0 0 -0.551804 0 0 0 0 0 1 0 0 1 0 0 479 1 0 0 0 0 0 -0.551804 0 0 0 0 0 0 0 0 480 481 rows × 20 columns final df = df clustered.copy() old_df = pd.read_excel('Asansol socio-economic data 1.xlsx') age = list(old df['Age (Yrs)']) final df['Age (Yrs)'] = age final df = final df.drop(columns = ['Normalised Age (Yrs)']) list2 = list(final df.columns) new order = [list2[-1]] + list2[0:-1]final_df = final_df[new_order] final df Out [27]: **Education Education** Income Income Income Helmet Age Gender Income Experience Experience Experience Experience Experience Experience **Higher Income** (Full- Clus (<Class (Class Graduate 20-40-60-<20k (20-30yr)(Yrs) (Male) Edu >80k (<1yr) (1-2yr) (2-5yr)(5-10yr) (10-20yr) (>30yr) 40k 80k 10) 10-12) 60k mask) 0 26 1 0 0 0 1 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 40 0 0 0 0 0 0 0 0 0 2 28 0 0 0 0 0 0 0 0 0 1 0 0 1 1 1 0 1 0 0 0 0 0 1 0 0 0 3 0 0 0 1 0 0 0 0 44 0 0 18 0 0 0 0 0 0 0 0 4 1 0 0 0 1 • • • 28 0 0 0 0 0 0 0 0 0 0 0 476 1 0 1 0 0 0 0 0 0 1 0 0 0 22 1 0 0 0 0 0 477 1 0 0 0 58 0 0 0 0 0 0 0 0 0 0 478 1 0 0 0 0 0 0 479 30 1 0 0 0 0 0 0 0 0 0 30 1 0 0 0 0 0 0 0 0 0 0 0 1 480 481 rows × 20 columns cluster 1 = final df.loc[final df['Cluster']==0] cluster 1 cluster 1.describe() Out[32]: Education Education Gender Higher **Experience Experience Experience Experience Experience** Income Income Income Income Income (<Class Age (Yrs) (Class Graduate Edu <20k 20-40k 40-60k 60-80k >80k (20-30yr)(Male) (<1yr) (1-2yr) (2-5yr) (5-10yr) (10-20yr) 10) 10-12) 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.000000 71.0 71.000000 71.000000 71.000000 71.000000 71.000000 count 51.873239 0.943662 0.140845 0.183099 0.577465 0.098592 0.070423 0.239437 0.267606 0.380282 0.042254 0.0 0.014085 0.028169 0.056338 0.169014 0.591549 mean 0.389500 0.300235 0.257679 0.429777 0.202599 0.232214 0.377432 0.495046 8.349388 0.232214 0.350338 0.497479 0.445862 0.488911 0.0 0.118678 0.166633 std **min** 35.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 46.000000 1.000000 0.000000 0.000000 0.0 51.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 1.000000 50% 1.000000 0.000000 0.0 **75%** 58.500000 1.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 1.000000 1.000000 0.000000 0.0 0.000000 0.000000 0.000000 0.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 **max** 73.000000 1.000000 0.0 cluster 2 = final df.loc[final df['Cluster']==1] cluster 2.describe() Out[30]: **Education Education** Gender Higher **Experience Experience Experience Experience Experience Experience** Income Income Income Income Income Age (Yrs) (<Class (Class 10-Graduate <20k 60-80k (Male) Edu 20-40k 40-60k >80k (20-30)(<1yr) (1-2yr) (2-5yr)(5-10yr) (10-20yr) 10) 12) 82.000000 82.000000 82.000000 82.000000 82.000000 82.000000 82.000000 82.000000 82.000 count 82.000000 82.000000 82.000000 82.000000 82.000000 82.000000 82.0 38.329268 0.951220 0.085366 0.878049 0.036585 0.073171 0.073171 0.658537 0.170732 0.024390 0.0 0.048780 0.048780 0.024390 0.853659 0.024 0.0 mean 0.378590 6.299235 0.216734 0.281145 0.329243 0.188897 0.262019 0.262019 0.477119 0.155207 0.0 0.216734 0.216734 0.155207 0.355623 0.155 0.0 std 28.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000 0.0 25% 35.000000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 0.000000 0.000000 0.000000 1.000000 0.000 0.0 0.000000 0.000000 0.000000 50% 37.000000 1.000000 0.0 0.000000 1.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.0 0.000000 1.000000 0.000 **75%** 40.750000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 1.000000 0.000 0.0 0.0 **max** 58.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 0.0 1.000000 1.000000 1.000000 1.000000 1.000 0.0 cluster 3 = final df.loc[final df['Cluster']==2] cluster 3.describe() Out[31]: **Education Education** Higher Income Experience Experience Experience Experience Experience Gender Income Income Income Income Age (Yrs) (<Class (Class 10-Graduate 40-60k 60-80k >80k (20-30yr)(Male) Edu <20k 20-40k (<1yr) (1-2yr) (2-5yr)(5-10yr) (10-20yr) 10) 12) 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.000000 97.0 97.00000 97.000000 97.000000 97.000000 32.298969 0.958763 0.742268 0.195876 0.030928 0.030928 0.865979 0.092784 0.020619 0.020619 0.051546 0.103093 0.14433 0.494845 0.175258 0.030928 0.0 0.174022 8.214780 0.199871 0.439658 0.398935 0.174022 0.174022 0.342444 0.291636 0.142842 0.142842 0.0 0.22258 0.305660 0.35325 0.502571 0.382162 std 0.000000 18.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.00000 0.000000 0.000000 0.000000 0.0 0.000000 0.000000 25% 26.000000 1.000000 0.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.0 0.000000 0.000000 0.00000 0.000000 0.000000 0.000000 32.000000 0.000000 1.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.0 0.000000 0.000000 0.00000 0.000000 0.000000 38.000000 1.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.00000 1.000000 0.000000 0.000000 0.0 **max** 50.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.00000 1.000000 1.000000 1.000000 cluster_4 = final_df.loc[final_df['Cluster']==3] cluster 4.describe() Out[33]: **Education Education** Income Income Experience Experience Experience Experience Experience Gender Income Income Income (<Class (Class 10-Age (Yrs) **Graduate Higher Edu** (Male) 60-80k (<1yr) 12) 60k 114.000000 114.000000 114.000000 114.000000 114.000000 114.000000 114.000000 **count** 114.000000 114.000000 114.000000 114.000000 114.000000 114.000000 114.0 114.000000 114.(35.350877 0.964912 0.061404 0.894737 0.017544 0.026316 0.070175 0.912281 0.0 0.008772 0.008772 0.052632 0.149123 0.096491 0.263158 0.412281 0. mean 9.370432 0.184814 0.241129 0.308247 0.131866 0.160779 0.256570 0.284135 0.093659 0.093659 0.224283 0.357782 0.296567 0.442292 0.494418 0. 0.0 std 16.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 0.0 min 25% 28.000000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 0.0 50% 34.500000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.0 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 **75%** 42.000000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 1.000000 1.000000 0.0 0.0 59.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.(0.0 max cluster 5 = final df.loc[final df['Cluster']==4] cluster 5.describe() Out[34]: **Education Education** Higher Gender Income Income Income Income Income Experience Experience Experience Experience Experience Exp Age (Yrs) (<Class (Class 10-Graduate Edu <20k 60-80k >80k (2-5yr)(Male) 20-40k 40-60k (<1yr) (1-2yr)(5-10yr) (10-20yr)10) 12) 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.000000 117.0 0.051282 0.042735 0.0 27.931624 0.931624 0.017094 0.888889 0.111111 0.299145 0.495726 0.068376 0.025641 0.034188 0.059829 0.282051 0.623932 mean 4.872029 0.253476 0.130179 0.221521 0.315621 0.203129 0.459853 0.502132 0.253476 0.158742 0.182493 0.238190 0.451934 0.486481 0.0 0.315621 std 0.000000 17.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 25.000000 1.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0 25% 0.000000 0.000000 50% 28.000000 1.000000 0.000000 0.000000 1.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 1.000000 0.0 30.000000 1.000000 0.000000 0.000000 1.000000 0.000000 0.000000 1.000000 1.000000 0.000000 0.000000 0.000000 0.000000 1.000000 1.000000 0.0 **75%** max 45.000000 1.000000 1.000000 0.0 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 1.000000 overall mean = final df.groupby('Cluster').mean() overall_mean Out[37]: **Education Education** Higher Experience Experience Experience Experience Experience Exp Gender Income Income Income Income Income Age (Yrs) (<Class (Class Graduate (Male) Edu <20k 20-40k 40-60k 60-80k >80k (1-2yr)(2-5yr)(10-20yr) (20-30yr) (<1yr) (5-10yr) 10) 10-12) Cluster 0.239437 0.014085 0.056338 51.873239 0.943662 0.140845 0.183099 0.577465 0.098592 0.070423 0.267606 0.380282 0.042254 0.000000 0.028169 0.169014 0.591549 0. **1** 38.329268 0.951220 0.000000 0.085366 0.878049 0.036585 0.073171 0.073171 0.658537 0.170732 0.024390 0.000000 0.048780 0.048780 0.024390 0.853659 0.024390 0. **2** 32.298969 0.958763 0.742268 0.195876 0.030928 0.030928 0.865979 0.092784 0.020619 0.020619 0.000000 0.051546 0.103093 0.144330 0.494845 0.175258 0.030928 0. 35.350877 0.964912 0.061404 0.894737 0.017544 0.026316 0.912281 0.000000 0.008772 0.052632 0.149123 0.096491 0.263158 0.412281 0.017544 0. 0.070175 0.008772 0.623932 27.931624 0.017094 0.051282 0.888889 0.042735 0.111111 0.299145 0.495726 0.068376 0.034188 0.059829 0.282051 0.000000 0.000000 0.931624 0.025641 In [40]: # Exporting the datasets to excel: final_df.to_excel('clustered data KMedoids.xlsx') overall mean.to excel('mean for clusters kmedoids.xlsx') cluster_1.to_excel('cluster_1_kmedoids.xlsx')

cluster_2.to_excel('cluster_2_kmedoids.xlsx')
cluster_3.to_excel('cluster_3_kmedoids.xlsx')
cluster_4.to_excel('cluster_4_kmedoids.xlsx')
cluster_5.to excel('cluster_5_kmedoids.xlsx')