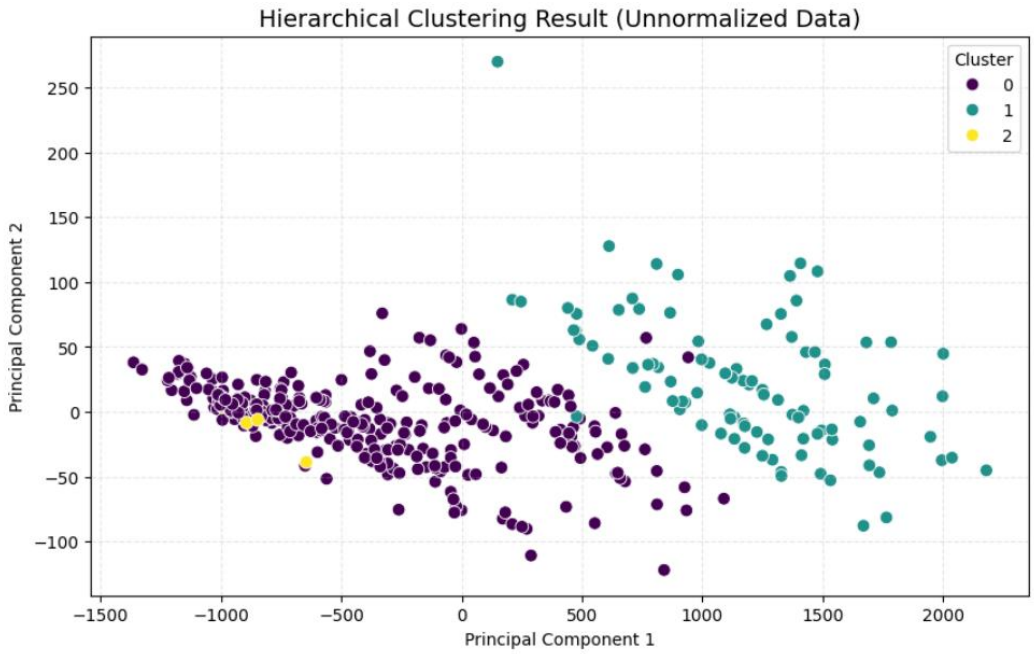
Problem 1

From the Jupyter Notebook analysis, all samples of origin 3 and the vast majority of origin 2 are assigned to cluster 0, indicating a strong connection between these two classes and cluster 0. Samples from origin 1 are distributed across cluster 0 and cluster 1, with the latter containing only a portion of this class's samples. Although certain classes exhibit clear associations, the clustering results do not form a clear one-to-one correspondence with the original class labels overall—primarily because cluster 0 incorporates samples from multiple classes, and cluster 2 includes only a negligible number of origin 2 samples. While some classes show localized strong correlations with specific clusters, the mixed composition of cluster 0 and the dispersed distribution of origin 1 samples prevent complete grouping according to the original classes.

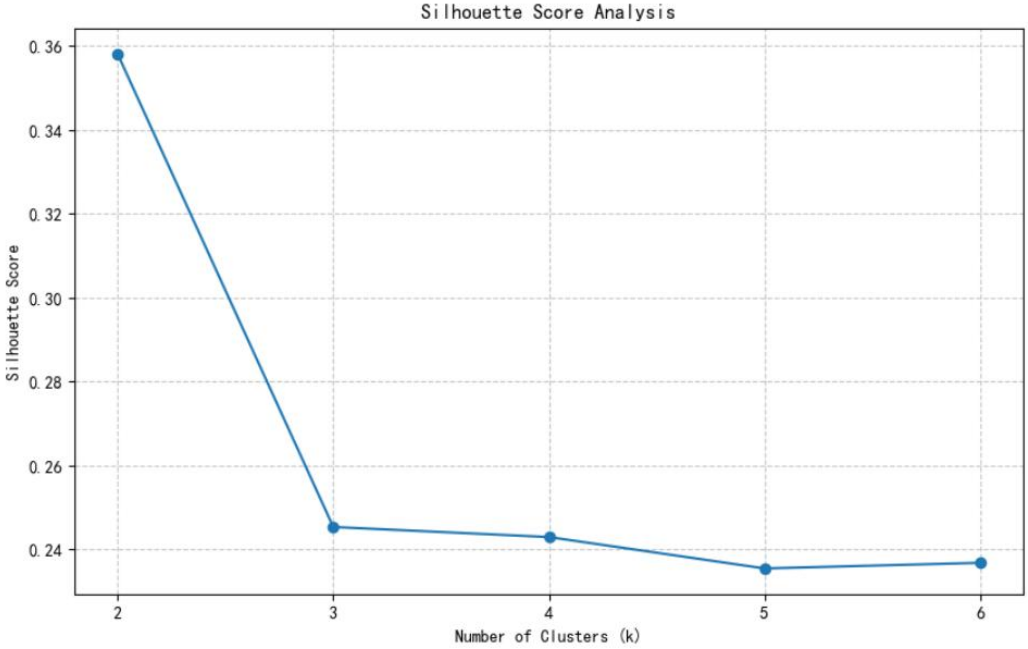
```
[Cross-Tabulation of Origin vs Cluster]:
Cluster    0   1  2  All
origin
1        152  97  0  249
2         66   0  4   70
3         79   0  0   79
All      297  97  4  398
```



Problem 2

From the Jupyter Notebook analysis， When k=2, the silhouette score reaches its highest value , indicating the optimal clustering effect with compact within-cluster structures and better separation between clusters. The silhouette scores for other k values (3–6) decrease significantly, suggesting that increasing the number of clusters does not lead to a clearer separation. Since the data is naturally divided into two categories,k=2 is the most reasonable choice.

Computational results show that after standardization, the mean values of all features for each cluster in the optimal clustering align exactly with the centroid coordinates, with negligible discrepancies originating from floating-point precision limitations.



Feature means for each cluster (original data):

	UNNAMED: 0	CRIM	ZN	INDUS	CHAS	NOX	RM
0	193.620896	0.287682	17.164179	7.178179	0.068657	0.489041	6.448764
1	370.807018	10.129061	0.000000	18.891930	0.070175	0.683316	5.963094

	AGE	DIS	RAD	TAX	PTRATIO	B
0	57.049552	4.710233	4.459701	302.480597	17.794030	384.797612
1	91.153801	2.002125	19.520468	615.421053	19.751462	301.578129

	LSTAT
0	9.519254
1	18.792398

Centroid coordinates for each cluster (inverse-transformed):

	UNNAMED: 0	CRIM	ZN	INDUS	CHAS	NOX
0	193.620896	0.287682	1.716418e+01	7.178179	0.068657	0.489041
1	370.807018	10.129061	1.243450e-14	18.891930	0.070175	0.683316

	RM	AGE	DIS	RAD	TAX	PTRATIO
0	6.448764	57.049552	4.710233	4.459701	302.480597	17.794030
1	5.963094	91.153801	2.002125	19.520468	615.421053	19.751462

	B	LSTAT
0	384.797612	9.519254
1	301.578129	18.792398

Differences between means and centroids:

	UNNAMED: 0	CRIM	ZN	INDUS	CHAS
0	0.000000e+00	1.110223e-15	-3.552714e-15	-2.664535e-15	8.326673e-17
1	5.684342e-14	-5.329071e-15	-1.243450e-14	-3.552714e-15	5.551115e-17

	NOX	RM	AGE	DIS	RAD	TAX
0	0.000000e+00	-8.881784e-16	-7.105427e-15	0.0	1.776357e-15	-5.684342e-14
1	-1.110223e-16	0.000000e+00	-1.421085e-14	0.0	-7.105427e-15	-6.821210e-13

	PTRATIO	B	LSTAT
0	0.000000e+00	5.684342e-14	0.000000e+00
1	-3.552714e-15	0.000000e+00	-3.552714e-15

Problem 3

Homogeneity measures whether each cluster contains samples from a single class. A value closer to 1 indicates that samples within each cluster belong to the same class, reflecting better clustering performance.Completeness measures whether all samples of a given class are assigned to the same cluster. A value closer to 1 indicates that samples of the same class are grouped together, reflecting better clustering performance.

In this dataset, both metrics approaching 1 suggest that the clustering results closely align with the true class labels.

```
Number of clusters: k=3
Homogeneity: 0.8788
Completeness: 0.8730
```