

clustering

November 21, 2024

```
[1]: import pandas as pd
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
data = pd.read_excel('/content/EastWestAirlines.xlsx', sheet_name='data')
data
```

```
[1]:
```

	ID#	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	1	28143	0	1	1	1	174	
1	2	19244	0	1	1	1	215	
2	3	41354	0	1	1	1	4123	
3	4	14776	0	1	1	1	500	
4	5	97752	0	4	1	1	43300	
...		
3994	4017	18476	0	1	1	1	8525	
3995	4018	64385	0	1	1	1	981	
3996	4019	73597	0	3	1	1	25447	
3997	4020	54899	0	1	1	1	500	
3998	4021	3016	0	1	1	1	0	

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
0	1	0	0	7000	
1	2	0	0	6968	
2	4	0	0	7034	
3	1	0	0	6952	
4	26	2077	4	6935	
...	
3994	4	200	1	1403	
3995	5	0	0	1395	
3996	8	0	0	1402	
3997	1	500	1	1401	
3998	0	0	0	1398	

	Award?
0	0
1	0
2	0

```

3          0
4          1
...      ...
3994       1
3995       1
3996       1
3997       0
3998       0

```

[3999 rows x 12 columns]

```
[2]: data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3999 entries, 0 to 3998
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID#                    3999 non-null   int64
1   Balance                3999 non-null   int64
2   Qual_miles             3999 non-null   int64
3   cc1_miles              3999 non-null   int64
4   cc2_miles              3999 non-null   int64
5   cc3_miles              3999 non-null   int64
6   Bonus_miles            3999 non-null   int64
7   Bonus_trans            3999 non-null   int64
8   Flight_miles_12mo      3999 non-null   int64
9   Flight_trans_12        3999 non-null   int64
10  Days_since_enroll      3999 non-null   int64
11  Award?                 3999 non-null   int64
dtypes: int64(12)
memory usage: 375.0 KB

```

```
[3]: data.isna().sum()
```

```

[3]: ID#          0
Balance          0
Qual_miles       0
cc1_miles        0
cc2_miles        0
cc3_miles        0
Bonus_miles      0
Bonus_trans      0
Flight_miles_12mo 0
Flight_trans_12  0
Days_since_enroll 0
Award?           0

```

dtype: int64

```
[4]: data.describe()
```

```
[4]:
```

	ID#	Balance	Qual_miles	cc1_miles	cc2_miles	\
count	3999.000000	3.999000e+03	3999.000000	3999.000000	3999.000000	
mean	2014.819455	7.360133e+04	144.114529	2.059515	1.014504	
std	1160.764358	1.007757e+05	773.663804	1.376919	0.147650	
min	1.000000	0.000000e+00	0.000000	1.000000	1.000000	
25%	1010.500000	1.852750e+04	0.000000	1.000000	1.000000	
50%	2016.000000	4.309700e+04	0.000000	1.000000	1.000000	
75%	3020.500000	9.240400e+04	0.000000	3.000000	1.000000	
max	4021.000000	1.704838e+06	11148.000000	5.000000	3.000000	

	cc3_miles	Bonus_miles	Bonus_trans	Flight_miles_12mo	\
count	3999.000000	3999.000000	3999.000000	3999.000000	
mean	1.012253	17144.846212	11.60190	460.055764	
std	0.195241	24150.967826	9.60381	1400.209171	
min	1.000000	0.000000	0.00000	0.000000	
25%	1.000000	1250.000000	3.00000	0.000000	
50%	1.000000	7171.000000	12.00000	0.000000	
75%	1.000000	23800.500000	17.00000	311.000000	
max	5.000000	263685.000000	86.00000	30817.000000	

	Flight_trans_12	Days_since_enroll	Award?
count	3999.000000	3999.000000	3999.000000
mean	1.373593	4118.55939	0.370343
std	3.793172	2065.13454	0.482957
min	0.000000	2.00000	0.000000
25%	0.000000	2330.00000	0.000000
50%	0.000000	4096.00000	0.000000
75%	1.000000	5790.50000	1.000000
max	53.000000	8296.00000	1.000000

```
[5]: data['ID#'].nunique()
```

```
[5]: 3999
```

```
[6]: data.drop(columns='ID#', inplace=True)
```

```
[7]: data.head()
```

```
[7]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	
3	14776	0	1	1	1	500	

4	97752	0	4	1	1	43300
	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	Award?	
0	1	0	0	7000	0	
1	2	0	0	6968	0	
2	4	0	0	7034	0	
3	1	0	0	6952	0	
4	26	2077	4	6935	1	

```
[8]: from sklearn.preprocessing import StandardScaler,MinMaxScaler
min_max_scaler = MinMaxScaler()
norm = min_max_scaler.fit_transform(data)
norm
```

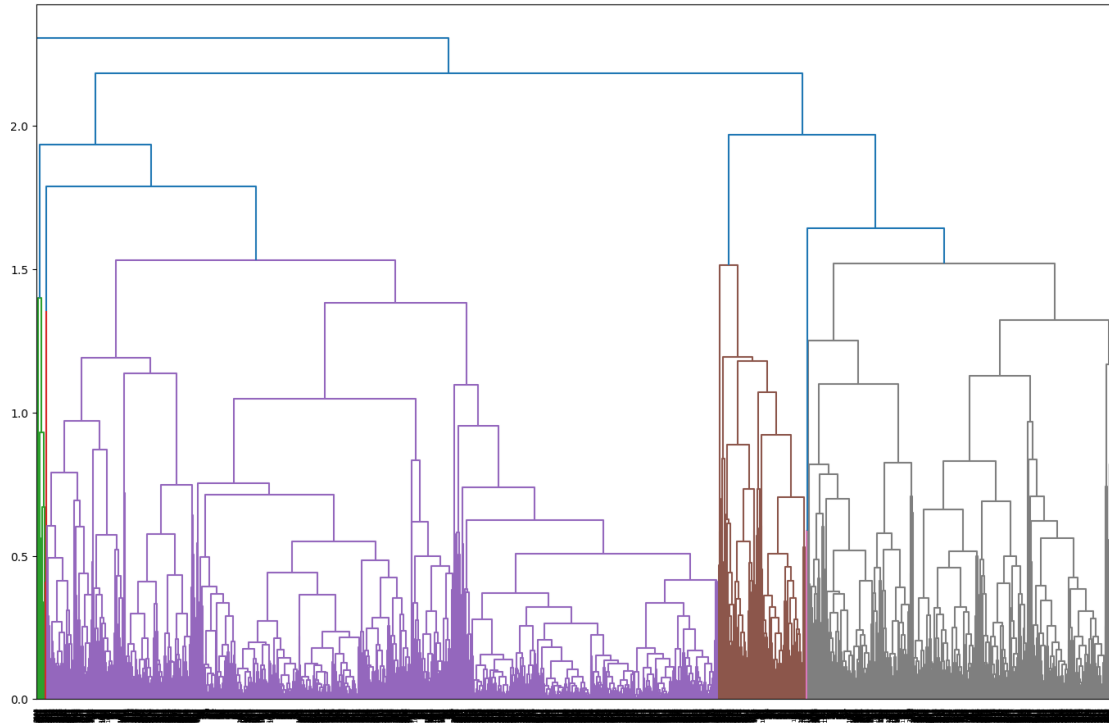
```
[8]: array([[0.01650773, 0.          , 0.          , ..., 0.          , 0.84374246,
0.          ],
[0.01128788, 0.          , 0.          , ..., 0.          , 0.83988425,
0.          ],
[0.02425685, 0.          , 0.          , ..., 0.          , 0.84784181,
0.          ],
...,
[0.0431695 , 0.          , 0.5          , ..., 0.          , 0.16879672,
1.          ],
[0.03220189, 0.          , 0.          , ..., 0.01886792, 0.16867615,
0.          ],
[0.00176908, 0.          , 0.          , ..., 0.          , 0.16831444,
0.          ]])
```

```
[9]: len(norm)
```

```
[9]: 3999
```

```
[10]: import scipy.cluster.hierarchy as sch
from sklearn.cluster import AgglomerativeClustering
```

```
[11]: plt.figure(figsize = (17,11))
dendrogram=sch.dendrogram(sch.linkage(norm, method='complete'))
```



```
[15]: hc = AgglomerativeClustering(n_clusters=5, linkage = 'complete')
      # Remove affinity parameter as it's not needed when linkage is 'complete'
      # For linkage='ward', affinity is fixed to 'euclidean' in older versions
      # If you need other distance metrics, consider upgrading scikit-learn
```

```
[16]: y_hc = hc.fit_predict(norm)
      y_hc
```

```
[16]: array([0, 0, 0, ..., 2, 0, 0])
```

```
[17]: hc.labels_
```

```
[17]: array([0, 0, 0, ..., 2, 0, 0])
```

```
[18]: data['h_clusterid'] = hc.labels_
```

```
[19]: data
```

```
[19]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	
3	14776	0	1	1	1	500	
4	97752	0	4	1	1	43300	

...
3994	18476	0	1	1	1	8525
3995	64385	0	1	1	1	981
3996	73597	0	3	1	1	25447
3997	54899	0	1	1	1	500
3998	3016	0	1	1	1	0

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
0	1	0	0	7000	
1	2	0	0	6968	
2	4	0	0	7034	
3	1	0	0	6952	
4	26	2077	4	6935	

...
3994	4	200	1	1403	
3995	5	0	0	1395	
3996	8	0	0	1402	
3997	1	500	1	1401	
3998	0	0	0	1398	

	Award?	h_clusterid
0	0	0
1	0	0
2	0	0
3	0	0
4	1	1

...
3994	1	2
3995	1	2
3996	1	2
3997	0	0
3998	0	0

[3999 rows x 12 columns]

```
[20]: data.h_clusterid.value_counts()
```

```
[20]: h_clusterid
0    2495
2    1144
1     325
4      31
3       4
Name: count, dtype: int64
```

```
[21]: data.sort_values("h_clusterid")
```

```
[21]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
2485	23649	0	1	1	1	3250	
2487	169794	0	3	1	1	22824	
2488	23781	0	1	1	1	50	
2491	5970	0	1	1	1	1000	
...	
940	44824	0	1	3	1	21107	
3959	13942	0	1	2	1	5822	
3779	35850	0	1	3	1	17759	
1389	49145	0	1	2	1	12755	
3191	56624	0	1	2	1	12311	

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
0	1	0	0	7000	
2485	16	0	0	3176	
2487	19	767	2	3119	
2488	1	50	1	3085	
2491	1	0	0	3091	
...	
940	19	2000	4	5916	
3959	12	0	0	1458	
3779	18	0	0	3439	
1389	22	2450	7	5323	
3191	14	0	0	2491	

	Award?	h_clusterid
0	0	0
2485	0	0
2487	0	0
2488	0	0
2491	0	0
...
940	1	4
3959	0	4
3779	0	4
1389	0	4
3191	0	4

[3999 rows x 12 columns]

```
[22]: data[data.h_clusterid==3]
```

```
[22]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
2015	53232	888	4	1	1	80696	
3235	287033	0	1	1	1	26161	
3583	160114	500	1	1	1	71954	

3594	27619	0	4	1	1	83726
------	-------	---	---	---	---	-------

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
2015	65	22100	45	3831	
3235	58	12873	53	2272	
3583	86	30817	53	1373	
3594	68	14050	46	1325	

	Award?	h_clusterid
2015	1	3
3235	1	3
3583	1	3
3594	1	3

```
[23]: data.groupby('h_clusterid').mean()
```

```
[23]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	\
h_clusterid						
0	59968.433667	88.883768	1.712224	1.000401	1.011222	
1	157084.578462	208.673846	4.661538	1.000000	1.061538	
2	80173.963287	248.550699	2.104895	1.009615	1.000874	
3	131999.500000	347.000000	2.500000	1.000000	1.000000	
4	45515.064516	32.258065	1.000000	2.483871	1.000000	

	Bonus_miles	Bonus_trans	Flight_miles_12mo	Flight_trans_12	\
h_clusterid					
0	10271.530261	9.105812	227.797194	0.656112	
1	70477.086154	23.249231	1019.433846	3.196923	
2	16882.864510	13.412587	739.958916	2.236888	
3	65634.250000	69.250000	19960.000000	49.250000	
4	14618.870968	16.129032	443.225806	1.967742	

	Days_since_enroll	Award?
h_clusterid		
0	3825.392786	0.000802
1	5770.572308	1.000000
2	4304.383741	1.000000
3	2200.250000	1.000000
4	3784.258065	0.193548

```
[24]: data2=data.drop(columns='h_clusterid')
data2.head()
```

```
[24]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	

3	14776	0	1	1	1	500
4	97752	0	4	1	1	43300

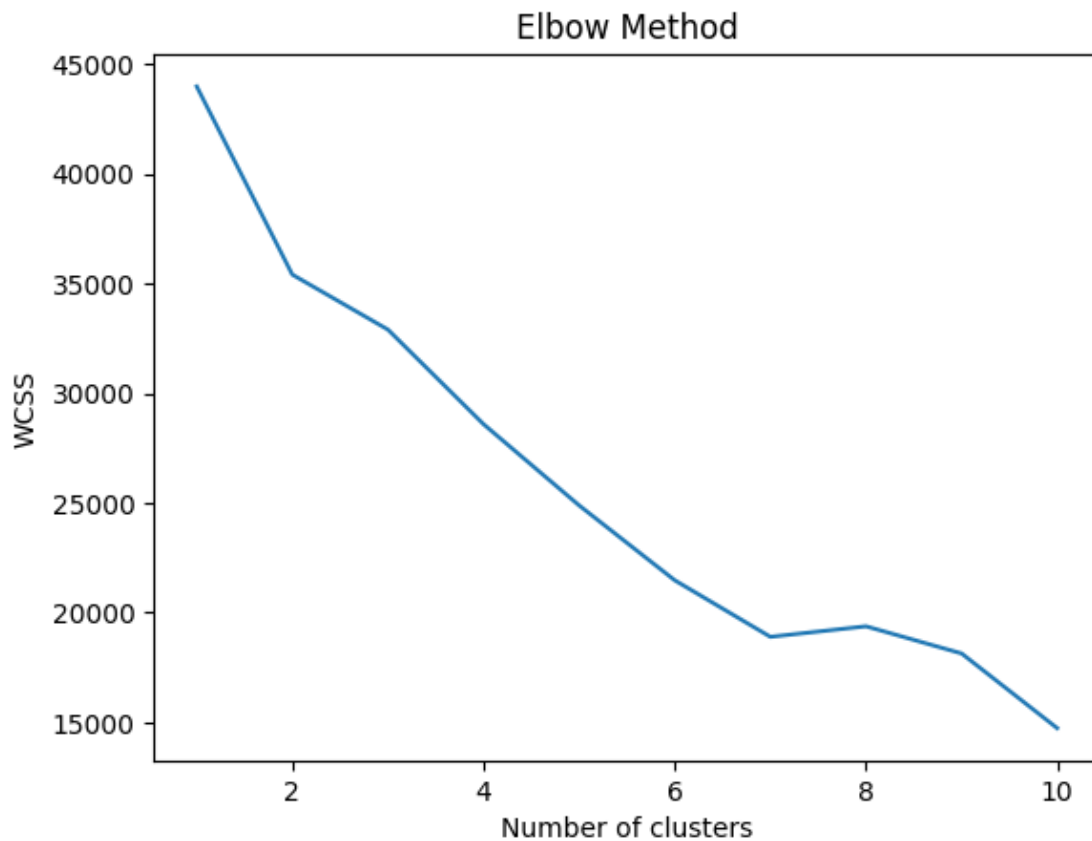
	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	Award?
0	1	0	0	7000	0
1	2	0	0	6968	0
2	4	0	0	7034	0
3	1	0	0	6952	0
4	26	2077	4	6935	1

```
[25]: from sklearn.cluster import KMeans
scaler = StandardScaler().fit_transform(data2)
scaler
```

```
[25]: array([[ -4.51140783e-01, -1.86298687e-01, -7.69578406e-01, ...,
        -3.62167870e-01,  1.39545434e+00, -7.66919299e-01],
        [ -5.39456874e-01, -1.86298687e-01, -7.69578406e-01, ...,
        -3.62167870e-01,  1.37995704e+00, -7.66919299e-01],
        [ -3.20031232e-01, -1.86298687e-01, -7.69578406e-01, ...,
        -3.62167870e-01,  1.41192021e+00, -7.66919299e-01],
        ...,
        [ -4.29480975e-05, -1.86298687e-01,  6.83121167e-01, ...,
        -3.62167870e-01, -1.31560393e+00,  1.30391816e+00],
        [ -1.85606976e-01, -1.86298687e-01, -7.69578406e-01, ...,
        -9.85033311e-02, -1.31608822e+00, -7.66919299e-01],
        [ -7.00507951e-01, -1.86298687e-01, -7.69578406e-01, ...,
        -3.62167870e-01, -1.31754109e+00, -7.66919299e-01]])
```

```
[26]: wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, random_state=0)
    kmeans.fit(scaler)
    wcss.append(kmeans.inertia_)

plt.plot(range(1, 11), wcss)
plt.title('Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```



```
[27]: wcss
```

```
[27]: [43989.0000000001,  
      35409.516629861384,  
      32901.45804218546,  
      28589.20585898687,  
      24884.115340435033,  
      21470.51454980924,  
      18891.750169018418,  
      19371.700292851994,  
      18134.257221538734,  
      14727.658831288609]
```

```
[28]: clusters = KMeans(6, random_state=42)  
clusters.fit(scaler)  
clusters.labels_
```

```
[28]: array([2, 2, 2, ..., 0, 3, 3], dtype=int32)
```

```
[29]: clusters.fit_predict(scaler)
```

```
[29]: array([2, 2, 2, ..., 0, 3, 3], dtype=int32)
```

```
[30]: clusters.inertia_
```

```
[30]: 25599.269402878585
```

```
[31]: from sklearn.metrics import silhouette_score
      silhouette_score(data2, clusters.labels_)
```

```
[31]: -0.03481925489833332
```

```
[32]: c2=KMeans(2, random_state=41)
      c2.fit(scaler)
```

```
[32]: KMeans(n_clusters=2, random_state=41)
```

```
[33]: silhouette_score(data2, c2.labels_)
```

```
[33]: 0.3446749880634653
```

```
[34]: data2['clusterid']=clusters.labels_
      data2
```

```
[34]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	
3	14776	0	1	1	1	500	
4	97752	0	4	1	1	43300	
...	
3994	18476	0	1	1	1	8525	
3995	64385	0	1	1	1	981	
3996	73597	0	3	1	1	25447	
3997	54899	0	1	1	1	500	
3998	3016	0	1	1	1	0	

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
0	1	0	0	7000	
1	2	0	0	6968	
2	4	0	0	7034	
3	1	0	0	6952	
4	26	2077	4	6935	
...	
3994	4	200	1	1403	
3995	5	0	0	1395	
3996	8	0	0	1402	
3997	1	500	1	1401	

3998	0	0	0	1398
------	---	---	---	------

	Award?	clusterid
0	0	2
1	0	2
2	0	2
3	0	2
4	1	1
...
3994	1	0
3995	1	0
3996	1	0
3997	0	3
3998	0	3

[3999 rows x 12 columns]

```
[35]: data2['clusterid'].value_counts()
```

```
[35]: clusterid
3    1185
2     980
0     819
1     800
4     128
5      87
Name: count, dtype: int64
```

```
[36]: data2.groupby('clusterid').mean()
```

```
[36]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	\
clusterid						
0	56127.758242	194.483516	1.675214	1.019536	1.000000	
1	107510.410000	136.801250	4.180000	1.001250	1.056250	
2	59727.241837	71.454082	1.481633	1.020408	1.001020	
3	34522.096203	93.091139	1.245570	1.011814	1.000000	
4	132067.828125	678.835938	2.132812	1.046875	1.000000	
5	528835.448276	463.931034	3.666667	1.011494	1.034483	

	Bonus_miles	Bonus_trans	Flight_miles_12mo	Flight_trans_12	\
clusterid					
0	10511.936508	10.586081	417.383394	1.249084	
1	48701.263750	19.910000	340.520000	1.022500	
2	6263.462245	8.532653	175.791837	0.517347	
3	4402.372996	6.589873	140.030380	0.422785	
4	30273.203125	28.906250	5995.664062	17.382812	
5	66229.632184	22.149425	1377.620690	4.816092	

	Days_since_enroll	Award?
clusterid		
0	4253.031746	1.00000
1	4817.865000	0.61125
2	5647.725510	0.00000
3	2103.353586	0.00000
4	4427.500000	0.78125
5	6191.137931	0.83908

```
[37]: from sklearn.cluster import DBSCAN
dbscan = DBSCAN(eps=1, min_samples=7)
dbscan.fit(scaler)
dbscan.labels_
```

```
[37]: array([0, 0, 0, ..., 1, 0, 0])
```

```
[38]: data3=data.drop(columns='h_clusterid')
data3.head()
```

```
[38]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	
3	14776	0	1	1	1	500	
4	97752	0	4	1	1	43300	

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	Award?
0	1	0	0	7000	0
1	2	0	0	6968	0
2	4	0	0	7034	0
3	1	0	0	6952	0
4	26	2077	4	6935	1

```
[39]: data3['clusterId']=dbscan.labels_
data3
```

```
[39]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	Bonus_miles	\
0	28143	0	1	1	1	174	
1	19244	0	1	1	1	215	
2	41354	0	1	1	1	4123	
3	14776	0	1	1	1	500	
4	97752	0	4	1	1	43300	
...	
3994	18476	0	1	1	1	8525	
3995	64385	0	1	1	1	981	
3996	73597	0	3	1	1	25447	

3997	54899	0	1	1	1	500
3998	3016	0	1	1	1	0

	Bonus_trans	Flight_miles_12mo	Flight_trans_12	Days_since_enroll	\
0	1	0	0	7000	
1	2	0	0	6968	
2	4	0	0	7034	
3	1	0	0	6952	
4	26	2077	4	6935	
...	
3994	4	200	1	1403	
3995	5	0	0	1395	
3996	8	0	0	1402	
3997	1	500	1	1401	
3998	0	0	0	1398	

	Award?	clusterId
0	0	0
1	0	0
2	0	0
3	0	0
4	1	1
...
3994	1	1
3995	1	1
3996	1	1
3997	0	0
3998	0	0

[3999 rows x 12 columns]

```
[40]: data3['clusterId'].value_counts()
```

```
[40]: clusterId
0      2299
1      1072
-1       604
2         11
4          8
3          5
Name: count, dtype: int64
```

```
[41]: data3.groupby('clusterId').mean()
```

```
[41]:
```

	Balance	Qual_miles	cc1_miles	cc2_miles	cc3_miles	\
clusterId						
-1	177887.369205	896.723510	2.690397	1.077815	1.081126	

0	52733.796433	3.679426	1.656807	1.000000	1.000000
1	60530.977612	6.554104	2.591418	1.000000	1.000000
2	28365.363636	0.000000	1.000000	2.000000	1.000000
3	51030.000000	0.000000	1.000000	1.000000	1.000000
4	24545.375000	2401.000000	1.000000	1.000000	1.000000

	Bonus_miles	Bonus_trans	Flight_miles_12mo	Flight_trans_12	\
clusterId					
-1	38686.490066	21.649007	2103.266556	6.089404	
0	9001.937364	8.428882	113.070465	0.373206	
1	22723.841418	12.793843	268.355410	0.840485	
2	8825.272727	11.818182	22.727273	0.181818	
3	4737.600000	11.400000	4242.800000	10.400000	
4	2427.750000	5.000000	37.500000	0.250000	

	Days_since_enroll	Award?
clusterId		
-1	4741.235099	0.677152
0	3791.968247	0.000000
1	4503.971082	1.000000
2	2702.000000	0.000000
3	2871.600000	0.000000
4	2042.625000	0.000000

[]: