```
1 import pandas as pd
 2 import numpy as np
 3 import matplotlib.pyplot as plt
 4 import seaborn as sns
 5 from sklearn.preprocessing import MinMaxScaler
 6 from sklearn.metrics import silhouette_score
 7 from sklearn.cluster import KMeans
8 pd.set_option('display.max_columns', None)
9 pd.set_option('display.max_rows', None)
10 pd.set_option('display.width', None)
11
12
13 class DataSet:
       def __init__(self, df):
14
15
           self.df = df
16
17
       def compute_statistical_summary(self):
           numerical_columns = self.df.select_dtypes(
18
   include=[np.number])
19
           if not numerical_columns.empty:
20
               numerical_stats = numerical_columns.agg
   (['mean', 'median', 'std'])
21
               print("Statistics for numerical columns
   :")
22
               print(numerical_stats)
23
           else:
               print("No numerical columns found.")
24
25
26
           categorical_columns = self.df.select_dtypes
   (include=['object', 'category'])
27
           if not categorical_columns.empty:
               print("\nFrequency distribution for
28
   categorical columns:")
29
               for column in categorical_columns:
30
                   freq_dist = ((self.df[column].
   value_counts() / len(df)) * 100).astype(str) + ' %'
31
                   print(f"\nColumn: {column}")
32
                   print(freq_dist)
33
34
       def visualize_data(self):
           sns.set_palette('bright')
35
```

```
36
           font_title = {'fontweight': 'bold',
   fontsize': 14}
37
           font_label = {'fontweight': 'bold', '
   fontsize': 12}
           numerical_columns = self.df.select_dtypes(
38
   include=[np.number])
39
           for column in numerical_columns:
40
               plt.figure()
41
               sns.histplot(self.df[column])
42
               plt.title(f"Histogram of {column}", **
   font_title)
43
               plt.xlabel(column, **font_label)
               plt.ylabel("Frequency", **font_label)
44
45
               plt.show()
46
           for column in numerical_columns:
47
               plt.figure()
48
               sns.boxplot(x=self.df[column])
               plt.title(f"Box plot of {column}", **
49
   font_title)
50
               plt.xlabel(column, **font_label)
               plt.ylabel("Value", **font_label)
51
52
               median_value = df[column].median()
               plt.text(0.5, 0.5, f"Median: {
53
   median_value:.2f}", transform=plt.gca().transAxes,
54
                        horizontalalignment='center',
   verticalalignment='center', **font_label)
55
               plt.show()
           sns.set(font_scale=1.2)
56
57
           sns.pairplot(self.df[numerical_columns.
   columns1)
58
           plt.title("Scatter Plot Matrix", **
   font_title)
59
           plt.show()
60
           categorical_columns = df.select_dtypes(
   include=['object', 'category'])
61
           for column in categorical_columns:
62
               plt.figure()
63
               sns.countplot(data=df, x=column)
               plt.title(f"Bar Graph of {column}", **
64
   font_title)
65
               plt.xlabel(column, **font_label)
```

```
plt.ylabel("Count", **font_label)
66
               value_counts = df[column].value_counts
67
   ()
               for i, count in enumerate(value_counts
68
   ):
69
                   plt.text(i, count, str(count), ha=
   'center', va='bottom', **font_label)
               plt.show()
70
71
72
       def preprocess_data(self):
           numerical_columns = self.df.select_dtypes(
73
   include=[np.number])
           scaler = MinMaxScaler()
74
75
           self.df[numerical columns.columns] =
   scaler.fit_transform(self.df[numerical_columns.
   columns])
           categorical_columns = self.df.
76
   select_dtypes(include=['object', 'category']).
   columns.tolist()
77
           self.df = pd.get_dummies(self.df, columns=
   categorical_columns)
78
79
       def find_optimal_clusters(self):
           X = self.df.drop('Customer_Id', axis=1)
80
81
           silhouette_scores = []
           k_range = range(2, 21)
82
83
           max_k = 2
           max\_score = -1
84
85
           for k in k_range:
               print(f"Running for k = {k}")
86
               kmeans = KMeans(n_clusters=k)
87
               kmeans.fit(X)
88
89
               labels = kmeans.predict(X)
               score = silhouette_score(X, labels)
90
91
               silhouette_scores.append(score)
92
               if score > max_score:
93
                   max_k = k
94
                   max_score = score
95
           print("Silhouette scores: ",
   silhouette_scores)
96
           return max_k
```

```
97
 98
        def perform_cluster_and_analyze(self,
    optimal_clusters):
            X = self.df.drop('Customer_Id', axis=1)
 99
            kmeans = KMeans(n_clusters=
100
    optimal_clusters)
            kmeans.fit(X)
101
102
            print("Number of customers in each cluster
    :", kmeans.labels_.size)
103
            cluster_labels = kmeans.labels_
104
            unique_labels, counts = np.unique(
    cluster_labels, return_counts=True)
            for cluster_label, count in zip(
105
    unique_labels, counts):
                print(f"Cluster {cluster_label}: {
106
    count} samples")
            X['Cluster'] = cluster_labels
107
            cluster_means = X.groupby('Cluster').mean
108
    ()
109
            print(cluster_means)
110
            cluster_means.to_csv('cluster_means.csv')
111
112
113 if __name__ == '__main__':
        df = pd.read_csv("JustBuy_data.csv")
114
115
        data_set = DataSet(df)
116
        data_set.compute_statistical_summary()
        data_set.visualize_data()
117
118
        data_set.preprocess_data()
        optimal_clusters = data_set.
119
    find_optimal_clusters()
        print("Optimal number of clusters: ",
120
    optimal_clusters)
121
        data_set.perform_cluster_and_analyze(
    optimal_clusters)
122
123
```