```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   from sklearn.cluster import KMeans
   from sklearn import preprocessing
   import mpl_toolkits

# load data
   df = pd.read_csv("cereal.csv")

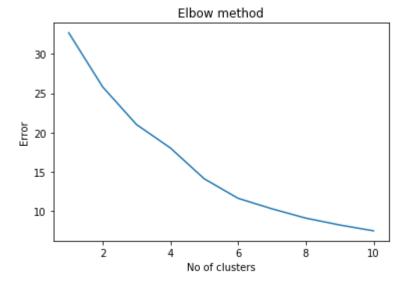
# Replace all '-1' with NaN and remove
   df.replace(-1, np.NaN, inplace=True)
   df.dropna(inplace = True)

In [2]: X = df[["salenias" "npotoin" "fat" "sadium" "fiben" "sanbo" "sugans" "pate
```

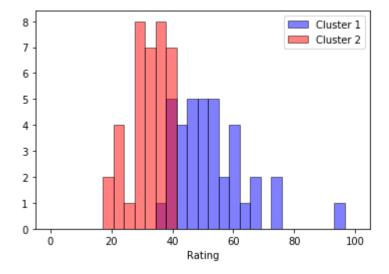
```
In [2]: X = df[["calories", "protein", "fat", "sodium", "fiber", "carbo", "sugars", "pota
```

```
In [3]: min_max_scaler = preprocessing.MinMaxScaler()
    norm_X = pd.DataFrame(min_max_scaler.fit_transform(X.values), columns = X.columns
```

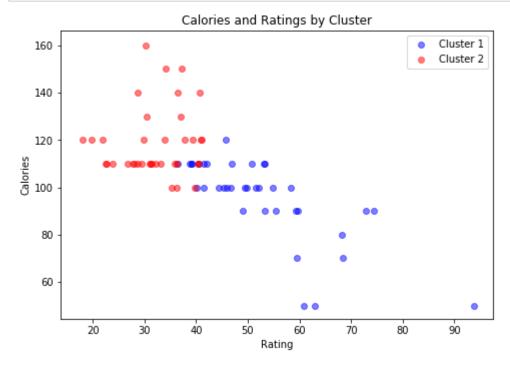
```
In [4]: Error =[]
    for i in range(1, 11):
        kmeans = KMeans(n_clusters = i).fit(norm_X)
        kmeans.fit(norm_X)
        Error.append(kmeans.inertia_)
    import matplotlib.pyplot as plt
    plt.plot(range(1, 11), Error)
    plt.title('Elbow method')
    plt.xlabel('No of clusters')
    plt.ylabel('Error')
    plt.show()
```



```
In [5]: |kmeans2 = KMeans(n clusters=2)
         y kmeans2 = kmeans2.fit predict(norm X)
         print(y_kmeans2)
         1001011101101100000110000010001001001
 In [6]: d = {'Cereal':df['name'], 'Cluster': y_kmeans2, 'Rating': df['rating'], 'Calories
         test = pd.DataFrame(data=d)
                                                                                       ▶
In [32]: test.head()
Out[32]:
               Cereal Cluster
                              Rating Calories Protein Fat Sodium Fiber Carbohydrates Sugars
               100%
                                         70
                                                     1
          0
                         0 68.402973
                                                 4
                                                          130
                                                               10.0
                                                                            5.0
                                                                                   6.0
                Bran
               100%
              Natural
                           33.983679
                                        120
                                                     5
                                                           15
                                                                2.0
                                                                            8.0
                                                                                   8.0
          1
                                                 3
                Bran
          2
              All-Bran
                         0 59.425505
                                         70
                                                     1
                                                          260
                                                                9.0
                                                                            7.0
                                                                                   5.0
              All-Bran
          3
            with Extra
                         0 93.704912
                                         50
                                                     0
                                                          140
                                                               14.0
                                                                            8.0
                                                                                   0.0
                Fiber
               Apple
            Cinnamon
                         1 29.509541
                                        110
                                                 2
                                                     2
                                                          180
                                                                            10.5
                                                                                  10.0
                                                                1.5
             Cheerios
 In [8]: |df['rating'].median()
 Out[8]: 40.2530865
 In [9]: | cluster 1 = test[test['Cluster'] == 0]
         med_1 = cluster_1['Rating'].median()
         mean_1 = cluster_1['Rating'].mean()
         print("Median of cluster 1: %f" %med 1)
         print("Mean of cluster 1: %f" %mean 1)
         Median of cluster 1: 50.764999
         Mean of cluster 1: 52.573632
         cluster 2 = test[test['Cluster'] == 1]
In [10]:
         med 2 = cluster 2['Rating'].median()
         mean_2 = cluster_2['Rating'].mean()
         print("Median of cluster 2: %f" %med 2)
         print("Mean of cluster 2: %f" %mean_2)
         Median of cluster 2: 32.207582
         Mean of cluster 2: 32.169942
```



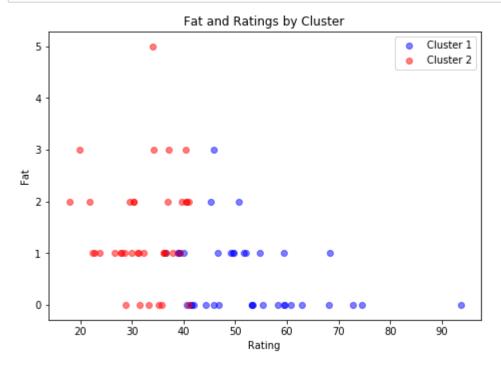
```
In [14]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Calories'], color='blue', alpha=0.5, ]
    ax.scatter(cluster_2['Rating'], cluster_2['Calories'], color='red', alpha=0.5, ]
    ax.set_xlabel('Rating')
    ax.set_ylabel('Calories')
    ax.set_title('Calories and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



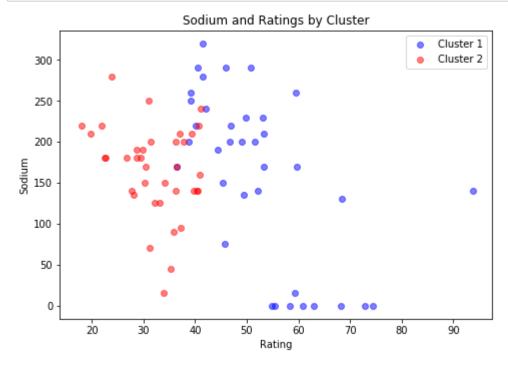
```
In [18]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Protein'], color='blue', alpha=0.5, latex.scatter(cluster_2['Rating'], cluster_2['Protein'], color='red', alpha=0.5, latex.set_xlabel('Rating')
    ax.set_ylabel('Protein')
    ax.set_title('Protein and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



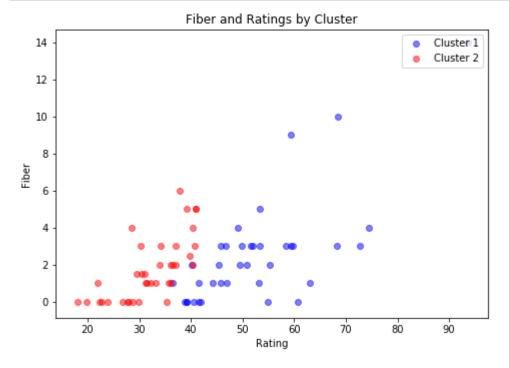
```
In [23]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Fat'], color='blue', alpha=0.5, label=
    ax.scatter(cluster_2['Rating'], cluster_2['Fat'], color='red', alpha=0.5, label='
    ax.set_xlabel('Rating')
    ax.set_ylabel('Fat')
    ax.set_title('Fat and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



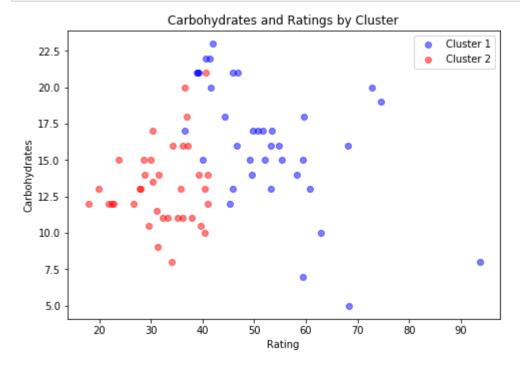
```
In [25]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Sodium'], color='blue', alpha=0.5, label ax.scatter(cluster_2['Rating'], cluster_2['Sodium'], color='red', alpha=0.5, label ax.set_xlabel('Rating')
    ax.set_ylabel('Sodium')
    ax.set_title('Sodium and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



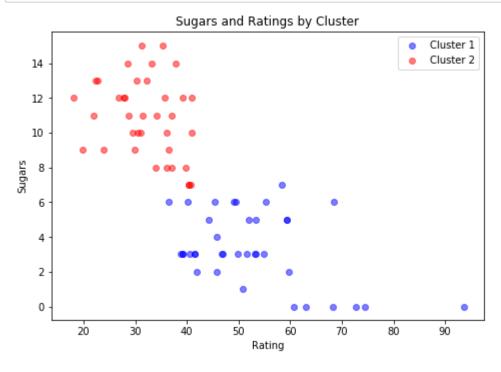
```
In [26]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Fiber'], color='blue', alpha=0.5, labe
    ax.scatter(cluster_2['Rating'], cluster_2['Fiber'], color='red', alpha=0.5, labe]
    ax.set_xlabel('Rating')
    ax.set_ylabel('Fiber')
    ax.set_title('Fiber and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



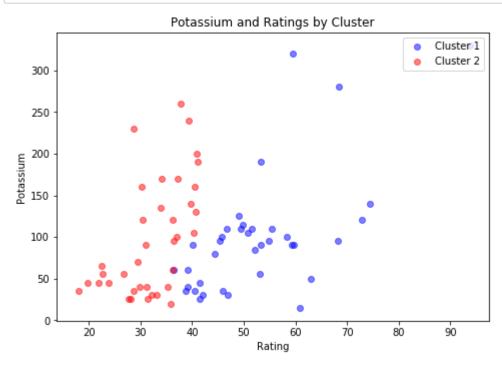
```
In [27]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Carbohydrates'], color='blue', alpha=@ax.scatter(cluster_2['Rating'], cluster_2['Carbohydrates'], color='red', alpha=@ax.set_xlabel('Rating')
    ax.set_ylabel('Carbohydrates')
    ax.set_title('Carbohydrates and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



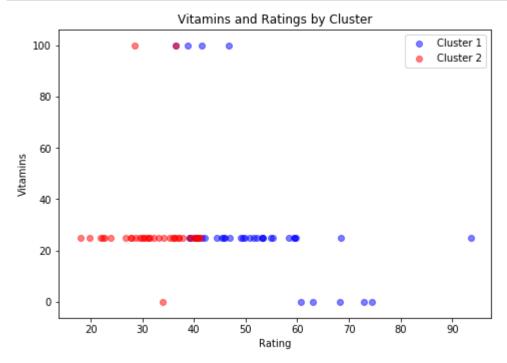
```
In [28]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Sugars'], color='blue', alpha=0.5, lak
    ax.scatter(cluster_2['Rating'], cluster_2['Sugars'], color='red', alpha=0.5, lake
    ax.set_xlabel('Rating')
    ax.set_ylabel('Sugars')
    ax.set_title('Sugars and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



```
In [29]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Potassium'], color='blue', alpha=0.5,
    ax.scatter(cluster_2['Rating'], cluster_2['Potassium'], color='red', alpha=0.5, ]
    ax.set_xlabel('Rating')
    ax.set_ylabel('Potassium')
    ax.set_title('Potassium and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



```
In [30]: fig=plt.figure()
    ax=fig.add_axes([0,0,1,1])
    ax.scatter(cluster_1['Rating'], cluster_1['Vitamins'], color='blue', alpha=0.5, later.scatter(cluster_2['Rating'], cluster_2['Vitamins'], color='red', alpha=0.5, later.scatter(later.scatter)
    ax.set_xlabel('Rating')
    ax.set_ylabel('Vitamins')
    ax.set_title('Vitamins and Ratings by Cluster')
    pyplot.legend(loc='upper right')
    plt.show()
```



```
In [34]: test_good = test.loc[test['Cluster'] == 0]
    test_bad = test.loc[test['Cluster'] == 1]
```

```
In [58]: # Sort by highest rating first
    test_good = test_good.sort_values('Rating', ascending=False)
    # Sort by Lowest rating first
    test_bad = test_bad.sort_values('Rating')
```

```
In [61]: print("TOP 10 GOOD FOR YOU\n", test good['Cereal'][0:10])
         print("\n")
         print("TOP 10 BAD FOR YOU\n", test_bad['Cereal'][0:10])
         TOP 10 GOOD FOR YOU
                All-Bran with Extra Fiber
          3
         64
                   Shredded Wheat 'n'Bran
         65
               Shredded Wheat spoon size
         0
                                100% Bran
                           Shredded Wheat
         63
         55
                             Puffed Wheat
                              Puffed Rice
         54
         50
                        Nutri-grain Wheat
         2
                                 All-Bran
         68
                  Strawberry Fruit Wheats
         Name: Cereal, dtype: object
         TOP 10 BAD FOR YOU
          10
                          Cap'n'Crunch
         12
               Cinnamon Toast Crunch
         35
                     Honey Graham Ohs
         18
                        Count Chocula
                          Cocoa Puffs
         14
                       Golden Grahams
         31
         42
                         Lucky Charms
         73
                                 Trix
         29
                       Fruity Pebbles
                    Total Raisin Bran
         Name: Cereal, dtype: object
```

In []: