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Batch-Roll number: C1-13

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import matplotlib.pyplot as plt \# for data visualization
import seaborn as sns # for statistical data visualization
%matplotlib inline
df = pd.read_csv('Live.csv')
df.drop(['Column1', 'Column2', 'Column3', 'Column4'], axis=1, inplace=True)
df.drop(['status_id', 'status_published'], axis=1, inplace=True)
X = df
y = df['status_type']
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
X['status_type'] = le.fit_transform(X['status_type'])
y = le.transform(y)
cols = X.columns
from sklearn.preprocessing import MinMaxScaler
ms = MinMaxScaler()
X = ms.fit_transform(X)
X = pd.DataFrame(X, columns=[cols])
X.head()
         status_type num_reactions num_comments num_shares num_likes num_loves num_wows num_hahas num_sads num_angrys
      0
            1.000000
                           0.112314
                                         0.024393
                                                     0.076519
                                                                0.091720
                                                                           0.140030 0.010791
                                                                                                0.006369
                                                                                                          0.019608
                                                                                                                           0.0
      1
            0.333333
                           0.031847
                                         0.000000
                                                     0.000000
                                                                0.031847
                                                                           0.000000
                                                                                     0.000000
                                                                                                0.000000
                                                                                                          0.000000
                                                                                                                           0.0
      2
            1.000000
                           0.048195
                                         0.011243
                                                     0.016647
                                                                0.043312
                                                                           0.031963
                                                                                     0.003597
                                                                                                0.006369
                                                                                                          0.000000
                                                                                                                           0.0
      3
            0.333333
                           0.023567
                                         0.000000
                                                     0.000000
                                                                0.023567
                                                                           0.000000
                                                                                     0.000000
                                                                                                0.000000
                                                                                                          0.000000
                                                                                                                           0.0
            0.333333
                           0.045223
                                         0.000000
                                                     0.000000
                                                                0.043312
                                                                           0.013699 0.000000
                                                                                                0.000000 0.000000
                                                                                                                           0.0
pip install sklearn.cluster
     ERROR: Could not find a version that satisfies the requirement sklearn.cluster (from versions: none)
     ERROR: No matching distribution found for sklearn.cluster
X.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 7050 entries, 0 to 7049
     Data columns (total 10 columns):
     #
         Column
                            Non-Null Count Dtype
     0
          (status_type,)
                            7050 non-null
                                            float64
          (num_reactions,)
                            7050 non-null
                                            float64
          (num_comments,)
                            7050 non-null
                                            float64
          (num_shares,)
                            7050 non-null
                                            float64
          (num_likes,)
                            7050 non-null
                                            float64
                            7050 non-null
                                            float64
          (num_loves,)
```

7050 non-null

float64

(num wows,)

```
7 (num_hahas,) 7050 non-null float64
8 (num_sads,) 7050 non-null float64
9 (num_angrys,) 7050 non-null float64
dtypes: float64(10)
```

memory usage: 550.9 KB

X.shape

(7050, 10)

X.describe()

	status_type	num_reactions	num_comments	num_shares	num_likes	num_loves	num_wows	num_hahas	num_sads	num_angr
count	7050.000000	7050.000000	7050.000000	7050.000000	7050.000000	7050.000000	7050.000000	7050.000000	7050.000000	7050.000
mean	0.568322	0.048857	0.010689	0.011689	0.045657	0.019374	0.004638	0.004436	0.004778	0.003
std	0.314133	0.098222	0.042384	0.038435	0.095429	0.060842	0.031366	0.025205	0.031317	0.023
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000
25%	0.333333	0.003609	0.000000	0.000000	0.003609	0.000000	0.000000	0.000000	0.000000	0.000
50%	0.333333	0.012633	0.000191	0.000000	0.012314	0.000000	0.000000	0.000000	0.000000	0.000
75%	1.000000	0.046497	0.001096	0.001168	0.039225	0.004566	0.000000	0.000000	0.000000	0.000
max	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000
4										<b>D</b>

from sklearn.cluster import KMeans

```
kmeans = KMeans(n_clusters=2, random_state=0)
```

kmeans.fit(X)

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/\_kmeans.py:870: FutureWarning: The default value of `n\_init` will change frow warnings.warn(

```
KMeans
KMeans(n_clusters=2, random_state=0)
```

```
kmeans.cluster_centers_
```

```
array([[3.28506857e-01, 3.90710874e-02, 7.54854864e-04, 7.53667113e-04, 3.85438884e-02, 2.17448568e-03, 2.43721364e-03, 1.20039760e-03, 2.75348016e-03, 1.45313276e-03], [9.54921576e-01, 6.46330441e-02, 2.67028654e-02, 2.93171709e-02, 5.71231462e-02, 4.71007076e-02, 8.18581889e-03, 9.65207685e-03, 8.04219428e-03, 7.19501847e-03]])
```

kmeans.inertia\_

237.75726404419646

labels = kmeans.labels\_

correct\_labels = sum(y == labels)
print("Result: %d out of %d samples were correctly labeled." % (correct\_labels, y.size))

Result: 63 out of 7050 samples were correctly labeled.

print('Accuracy score: {0:0.2f}'. format(correct\_labels/float(y.size)))

Accuracy score: 0.01

```
from sklearn.cluster import KMeans
cs = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', max_iter = 300, n_init = 10, random_state = 0)
    kmeans.fit(X)
    cs.append(kmeans.inertia_)
plt.plot(range(1, 11), cs)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('CS')
plt.show()
```

## The Elbow Method 800 - 600 - 200 -

```
from sklearn.cluster import KMeans
kmeans = KMeans(n_clusters=4,random_state=0)
kmeans.fit(X)
labels = kmeans.labels
# check how many of the samples were correctly labeled
correct_labels = sum(y == labels)
print("Result: %d out of %d samples were correctly labeled." % (correct_labels, y.size))
print('Accuracy score: {0:0.2f}'. format(correct_labels/float(y.size)))
                      /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from the control of the con
                              warnings.warn(
                      Result: 4340 out of 7050 samples were correctly labeled.
                     Accuracy score: 0.62
from sklearn.cluster import KMeans
kmeans = KMeans(n_clusters=3,random_state=0)
kmeans.fit(X)
labels = kmeans.labels_
# check how many of the samples were correctly labeled
correct_labels = sum(y == labels)
print("Result: %d out of %d samples were correctly labeled." % (correct_labels, y.size))
print('Accuracy score: {0:0.2f}'. format(correct_labels/float(y.size)))
                      /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from the control of the con
                             warnings.warn(
                      Result: 138 out of 7050 samples were correctly labeled.
                      Accuracy score: 0.02
```