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Branch-CSE(AIML) Year: 2021-2025

```
#dataset = https://www.kaggle.com/code/prashant111/k-means-clustering-with-
python?scriptVersionId=48823469&cellId=18
import pandas as pd
df = pd.read_csv('/content/Live.csv')
df.shape#7050 rows and 16 cols
df.info()
df.isnull().sum() #checking for missing values
df.describe()
# view the labels in the variable
df['status_id'].unique()
# view how many different types of variables are there
len(df['status_id'].unique())
# view the labels in the variable
df['status_published'].unique()
# view how many different types of variables are there
len(df['status_published'].unique())
# view the labels in the variable
df['status_type'].unique()
# view how many different types of variables are there
len(df['status_type'].unique())
df.drop(['status_id', 'status_published'], axis=1, inplace=True)
df.info()
df.drop(['Column1', 'Column2', 'Column3', 'Column4'], axis=1, inplace=True)
df.info()
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)
X.info()
#feature scaling
cols = X.columns
from sklearn.preprocessing import MinMaxScaler
ms = MinMaxScaler()
X = ms.fit_transform(X)
X = pd.DataFrame(X, columns=[cols])
X.head()
#ELBOW Method
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
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()
from sklearn.cluster import KMeans
kmeans = KMeans(n clusters=2,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)))
```

```
#with 4 clusters
kmeans = KMeans(n_clusters=4, random_state=0)
kmeans.fit(X)
# check how many of the samples were correctly labeled
labels = kmeans.labels_
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)))
#with 3 clusters
#K-Means model with 3 clusters
kmeans = KMeans(n_clusters=3, random_state=0)
kmeans.fit(X)
# check how many of the samples were correctly labeled
labels = kmeans.labels_
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)))
```

## **OUTPUTS:**

<>

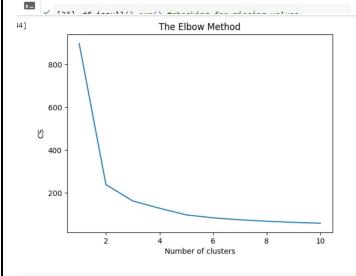
Column4

₽		status_id	status_type	status_published	num_reactions	num_comments	num_shares	num_likes	num_loves	num_wows	num_hahas	n
	0	246675545449582_1649696485147474	video	4/22/2018 6:00	529	512	262	432	92	3	1	
1 2 3 4  7045	1	246675545449582_1649426988507757	photo	4/21/2018 22:45	150	0	0	150	0	0	0	
	2	246675545449582_1648730588577397	video	4/21/2018 6:17	227	236	57	204	21	1	1	
	3	246675545449582_1648576705259452	photo	4/21/2018 2:29	111	0	0	111	0	0	0	
	4	246675545449582_1645700502213739	photo	4/18/2018 3:22	213	0	0	204	9	0	0	
	045	1050855161656896_1061863470556065	photo	9/24/2016 2:58	89	0	0	89	0	0	0	
	046	1050855161656896_1061334757275603	photo	9/23/2016 11:19	16	0	0	14	1	0	1	
70	047	1050855161656896_1060126464063099	photo	9/21/2016 23:03	2	0	0	1	1	0	0	
7048 7049	048	1050855161656896_1058663487542730	photo	9/20/2016 0:43	351	12	22	349	2	0	0	
	049	1050855161656896_1050858841656528	photo	9/10/2016 10:30	17	0	0	17	0	0	0	
705	50 rov	ws × 16 columns										

>\_ df.info() 0 Q <class 'pandas.core.frame.DataFrame'> RangeIndex: 7050 entries, 0 to 7049 {*x*} Data columns (total 16 columns): Column Non-Null Count Dtype # status\_id status\_type 0 7050 non-null object 7050 non-null object status\_published 7050 non-null object num\_reactions 7050 non-null int64 num\_comments 7050 non-null int64 7050 non-null 7050 non-null num\_shares int64 num\_likes int64 num\_loves 7050 non-null int64 num\_wows num\_hahas 7050 non-null int64 7050 non-null int64 num\_sads 7050 non-null int64 7050 non-null 11 num\_angrys int64 Column1 0 non-null float64 Column2 0 non-null float64 14 Column3 0 non-null float64 0 non-null

float64





dtypes: float64(4), int64(9), object(3) memory usage: 881.4+ KB

() /usr/local/lib/python3.10/dist-packages/sklearn/cluster/\_kmeans.py:870: FutureWarning: The default value of `n\_init` will change from 10 to 'auto' in 1.4. Set 1 warnings.warn( Wailings.wailing Result: 4340 out of 7050 samples were correctly labeled. Accuracy score: 0.62 {x} 

#K-Means model with 3 clusters

"Means(n clusters=3, t kmeans = KMeans(n\_clusters=3, random\_state=0) # check how many of the samples were correctly labeled labels = kmeans.labels\_ 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 10 to 'auto' in 1.4. Set t warnings.warn(
Result: 138 out of 7050 samples were correctly labeled. Accuracy score: 0.02 racking in the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the default will be also will be al warnings.warn( Result: 63 out of 7050 samples were correctly labeled. Accuracy score: 0.01