Week7_7_Kmeans_cluster_3d_git

May 21, 2021

K Means Clustering

Import Libraries

```
[1]: from sklearn.datasets import make_blobs
     from sklearn.datasets import make_gaussian_quantiles
     from sklearn.datasets import make_classification, make_regression
     import argparse
     import json
     import re
     import os
     import sys
     import plotly
     import plotly.graph_objs as go
     plotly.offline.init_notebook_mode()
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     df = pd.read_csv('/home/jayanthikishore/Desktop/Analysis/Work/ML_EIT/Data/

→Mall_Customers.csv')
     df.head()
```

[1]:	${\tt CustomerID}$	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

Data Shape and display specified columns

```
[2]: df.shape
```

```
[2]: (200, 5)
```

```
[3]: #specific columns only copy

df_sel = df.iloc[:,[2,3,4]].values

df_sel.shape
```

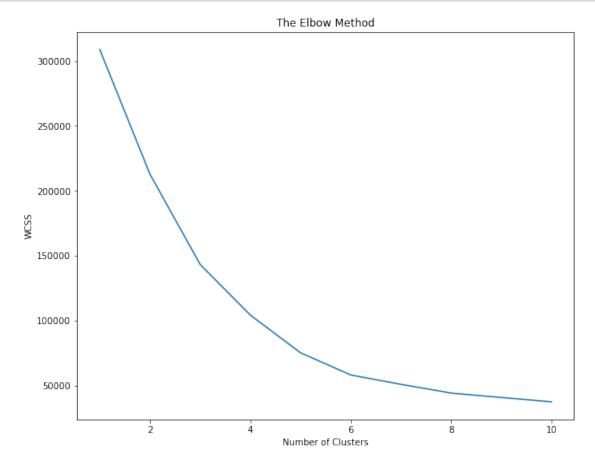
[3]: (200, 3)

Import K Means Cluster

```
[4]: from sklearn.cluster import KMeans

fig = plt.figure(figsize=(10, 8))
WCSS = []
for i in range(1, 11):
    clf = KMeans(n_clusters=i, init='k-means++', max_iter=300, n_init=10, \( \) \( \text{aradom_state=0} \)
    clf.fit(df_sel)
    WCSS.append(clf.inertia_) # inertia is another name for WCSS

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



Choosing n clusters

 $\bullet \ \ \textit{From ELBOW method the optimum value is 6 clusters}$

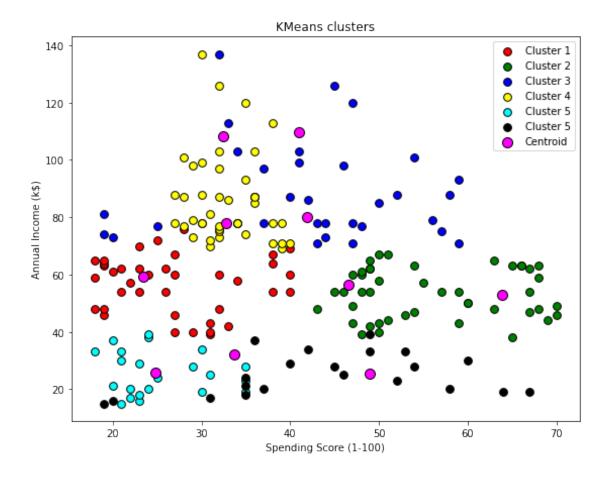
```
[5]: clsters = KMeans(n clusters=6, init='k-means++', max iter=300, n init=10, L
           →random_state=0)
         nclsters y = clsters.fit predict(df sel)
[6]: nclsters_y
[6]: array([5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 
                       5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 5, 4, 1, 4, 1, 0,
                       5, 4, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0,
                       1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0,
                       0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1,
                       1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 3, 0, 3, 2, 3, 2, 3, 2, 3,
                       0, 3, 2, 3, 2, 3, 2, 3, 2, 3, 0, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                       2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                       2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                       2, 3], dtype=int32)
[7]: fig = plt.figure(figsize=(9, 7))
         plt.scatter(df_sel[nclsters_y == 0, 0], df_sel[nclsters_y == 0, 1],__

→color='red', s=60, label='Cluster 1', edgecolors='black')
         plt.scatter(df_sel[nclsters_y == 1, 0], df_sel[nclsters_y == 1, 1],__

→color='green', s=60, label='Cluster 2', edgecolors='black')
         plt.scatter(df_sel[nclsters_y == 2, 0], df_sel[nclsters_y == 2, 1],__
           ⇒color='blue',s=60, label='Cluster 3', edgecolors='black')
         plt.scatter(df_sel[nclsters_y == 3, 0], df_sel[nclsters_y == 3, 1],_
           plt.scatter(df_sel[nclsters_y == 4, 0], df_sel[nclsters_y == 4, 1],__

→color='cyan', s=60, label='Cluster 5', edgecolors='black')
         plt.scatter(df sel[nclsters y == 5, 0], df sel[nclsters y == 5, 1],
           # cluster centres
         plt.scatter(clf.cluster_centers_[:, 0], clf.cluster_centers_[:, 1],_

→color='magenta', s=100, label='Centroid', edgecolors='black')
         plt.legend()
         plt.title('KMeans clusters')
         plt.ylabel('Annual Income (k$)')
         plt.xlabel('Spending Score (1-100)')
         plt.show()
```



```
[8]: #In this dataframe three columns are Age Annual Income (k$)

Spending Score (1-100)

#so, I am changing to common columns as x0, x1, and x2

def columns_rename(dff, prefix='x'):

dff = dff.copy()
 dff.columns = [prefix + str(i) for i in dff.columns]

return dff

# creating dataFrame of df_sel
 df_new = pd.DataFrame(df_sel)
 df_new.head(5)
```

```
[8]: 0 1 2
0 19 15 39
1 21 15 81
2 20 16 6
```

```
4 31 17 40
 [9]: #Rename column names
     df_new = colmns_rename(df_new)
     df_new.head(5)
 [9]:
        x0 x1
                x2
     0 19 15 39
     1
       21
           15 81
     2 20 16
                6
     3 23 16 77
     4 31
           17 40
[10]: #adding the nclsters_y as a y_columns
     df_new['yy'] = nclsters_y
     df_new.head()
[10]:
        x0
           x1
                x2
                    уу
     0 19
           15
                39
                    5
        21
            15 81
                     4
     2 20 16
                6
                     5
        23 16 77
                     4
     3
     4 31 17
               40
                     5
[11]: %run -i '~/Desktop/Analysis/Work/ML_EIT/Github/cluster3_3d.py'
     cluster3_3d(df_new)
[12]: %run -i '~/Desktop/Analysis/Work/ML_EIT/Github/clusters4_3d.py'
     clusters4_3d(df_new)
 []:
```

3 23 16 77