k-means

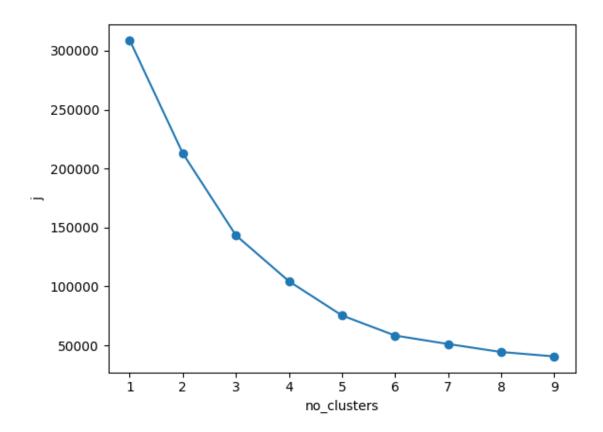
August 27, 2023

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     from sklearn.cluster import KMeans
     from sklearn.preprocessing import LabelEncoder
[2]: data=pd.read_csv('Mall_Customers.csv')
[3]: data
[3]:
          CustomerID
                      Gender
                               Age
                                    Annual Income (k$)
                                                         Spending Score (1-100)
                    1
                         Male
                                19
                                                     15
                                                                               39
     1
                    2
                         Male
                                21
                                                     15
                                                                               81
                                20
     2
                   3 Female
                                                     16
                                                                               6
     3
                   4
                      Female
                                23
                                                                               77
                                                     16
     4
                   5
                      Female
                                31
                                                     17
                                                                               40
     195
                 196
                      Female
                                35
                                                    120
                                                                               79
     196
                 197
                      Female
                                45
                                                    126
                                                                               28
     197
                 198
                         Male
                                32
                                                    126
                                                                               74
     198
                         Male
                                32
                                                                               18
                 199
                                                    137
     199
                 200
                         Male
                                30
                                                    137
                                                                               83
     [200 rows x 5 columns]
[4]: # Data Info :
     data.shape
[5]: (200, 5)
[6]: data.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200 entries, 0 to 199
    Data columns (total 5 columns):
         Column
                                   Non-Null Count Dtype
```

```
CustomerID
                                                      int64
      0
                                     200 non-null
      1
           Gender
                                    200 non-null
                                                      object
      2
           Age
                                    200 non-null
                                                      int64
           Annual Income (k$)
                                    200 non-null
                                                      int64
      3
           Spending Score (1-100)
                                    200 non-null
                                                      int64
     dtypes: int64(4), object(1)
     memory usage: 7.9+ KB
 [7]: data.describe()
 [7]:
             CustomerID
                                  Age
                                       Annual Income (k$)
                                                             Spending Score (1-100)
             200.000000
                          200.000000
                                                200.000000
                                                                          200.000000
      count
      mean
              100.500000
                           38.850000
                                                 60.560000
                                                                           50.200000
      std
              57.879185
                           13.969007
                                                 26.264721
                                                                           25.823522
      min
                1.000000
                            18.000000
                                                 15.000000
                                                                            1.000000
      25%
               50.750000
                            28.750000
                                                                           34.750000
                                                 41.500000
      50%
              100.500000
                           36.000000
                                                 61.500000
                                                                           50.000000
      75%
              150.250000
                           49.000000
                                                 78.000000
                                                                           73.000000
      max
              200.000000
                           70.000000
                                                137.000000
                                                                           99.000000
 [8]: data.drop(['CustomerID'],axis=1,inplace=True)
                                                                #Delete CustomerID from
        \rightarrow data
 [9]:
      data
 [9]:
           Gender
                    Age
                         Annual Income (k$)
                                               Spending Score (1-100)
      0
             Male
                     19
                                           15
                                                                    39
             Male
                                           15
                                                                    81
      1
                     21
      2
           Female
                     20
                                           16
                                                                     6
      3
           Female
                     23
                                           16
                                                                    77
      4
           Female
                     31
                                           17
                                                                    40
      . .
                                                                    79
      195
           Female
                     35
                                          120
      196
          Female
                     45
                                          126
                                                                    28
      197
                                          126
                                                                    74
             Male
                     32
      198
             Male
                     32
                                          137
                                                                    18
      199
             Male
                                                                    83
                     30
                                          137
      [200 rows x 4 columns]
[10]: la=LabelEncoder()
                                   #to convert object column to numerical
[11]: data['Gender']=la.fit_transform(data['Gender'])
[12]:
      data
```

[12]:	Gender	Age	Annual	Income	(k\$)	Spending	Score	(1-10	00)		
0	1	19			15				39		
1	1	21			15				81		
2	0	20			16				6		
3	0	23			16				77		
4	0	31			17				40		
				•••							
	95 0	35			120				79		
	96 0	45			126				28		
	97 1	32			126				74		
		32			137				18		
1	99 1	30			137				83		
	200 rows x	4 colı	umns]								
[13]: #	Apply the A	KMeans									
[14] · [n	[1/]: no alwatera=[] # to atoms the walker of alwaters										
	[14]: no_clusters=[] # to store the values of clusters j=[]										
[45] . [4	or i in ran	(4	10).								
	model=KM model.fi no_clust j.append	Means(1 Lt(data ters.a	n_clusto a) ppend(i))							
Fı 1. /ı	usr/local/litureWarning 4. Set the warnings.wa	g: The value arn(ib/pyt	defaul of `n_ hon3.10	t value init` e /dist-p	of `nexplic	n_init` wi itly to su es/sklearn	ll chan nppress	nge for the er/_k	rom 10 to warning means.py:8	'auto' 370:	
1.	tureWarning 4. Set the warnings.wa	value arn(of `n_	init` e	xplic	itly to su	ıppress	the	warning		in
Fu	usr/local/li utureWarning 4. Set the warnings.wa	g: The value	defaul	t value	of `ı	n_init` wi	ll chai	nge f	rom 10 to		in
Fu	usr/local/litureWarning 4. Set the warnings.wa	g: The value	defaul	t value	of `i	n_init` wi	ll cha	nge f	rom 10 to		in
Fu	usr/local/latureWarning 4. Set the warnings.wa	ib/pyt g: The value	defaul	t value	of `ı	n_init` wi	ll chai	nge f	rom 10 to		in
/۱	sr/local/l		hon3.10	/dist-p	ackage	es/sklearn	n/clust	er/_k	means.py:8	370:	

```
FutureWarning: The default value of `n_init` will change from 10 to 'auto' in
     1.4. Set the value of `n_init` explicitly to suppress the warning
       warnings.warn(
     /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 the value of `n_init` explicitly to suppress the warning
       warnings.warn(
     /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 the value of `n_init` explicitly to suppress the warning
       warnings.warn(
     /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 the value of `n_init` explicitly to suppress the warning
       warnings.warn(
[16]: pd.DataFrame(no_clusters, j)
                                       #make DataFrame contain (no_clusters, j)
[16]:
                     0
      308862.060000
      212889.442455 2
      143391.592360 3
      104414.675342 4
      75399.615414
      58348.641363
      51165.184237
                     7
      44391.820805
      40639.660395
[17]: plt.plot(no_clusters,j,marker='o')
      plt.xlabel('no_clusters')
      plt.ylabel('j')
[17]: Text(0, 0.5, 'j')
```



```
[19]: model=KMeans(n_clusters=5)
model.fit(data)
pre=model.predict(data)
```

/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 the value of `n_init` explicitly to suppress the warning
warnings.warn(

```
[20]: data['KMeans']=pre #add KMeans column to data
```

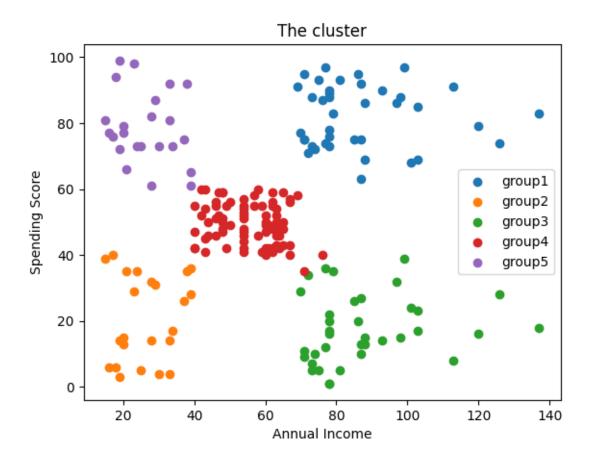
[21]: data

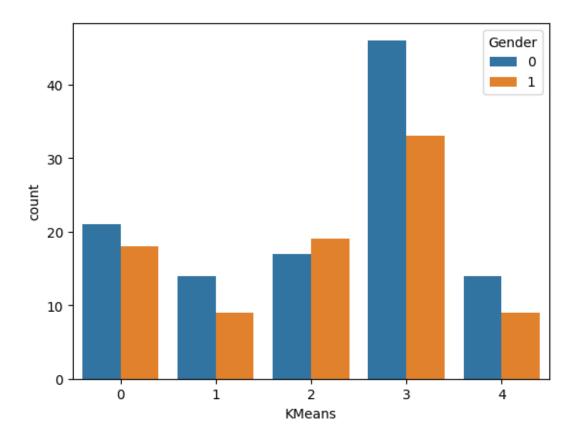
[21]:	Gender	Age	Annual Income (k\$)	Spending Score (1-100)	KMeans
0	1	19	15	39	1
1	1	21	15	81	4
2	0	20	16	6	1
3	0	23	16	77	4
4	0	31	17	40	1
			•••	•••	
195	0	35	120	79	0
196	0	45	126	28	2

```
197
                                   126
                                                          74
                                                                   0
              1
                  32
     198
                  32
                                   137
                                                           18
                                                                   2
              1
     199
                                                          83
                                                                   0
              1
                  30
                                   137
     [200 rows x 5 columns]
[22]: group1=data[data['KMeans']==0]
     group2=data[data['KMeans']==1]
     group3=data[data['KMeans']==2]
     group4=data[data['KMeans']==3]
     group5=data[data['KMeans']==4]
[23]: # Numerical Features vs Numerical Features w.r.t Categorical Feature
[24]: #plot the final cluster
[25]: plt.scatter(group1['Annual Income (k$)'],group1['Spending Score
      →(1-100)'],label='group1')
     plt.scatter(group2['Annual Income (k$)'],group2['Spending Score_
      ⇔(1-100)'],label='group2')
     plt.scatter(group3['Annual Income (k$)'],group3['Spending Score_
      plt.scatter(group4['Annual Income (k$)'],group4['Spending ScoreL
      plt.scatter(group5['Annual Income (k$)'],group5['Spending Score_
      plt.legend()
     plt.title('The cluster')
     plt.xlabel('Annual Income')
```

[25]: Text(0, 0.5, 'Spending Score')

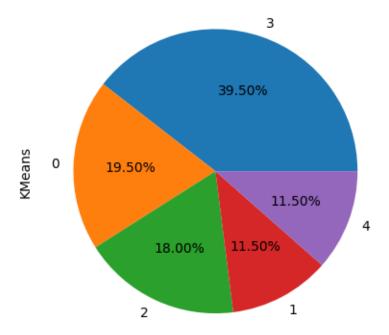
plt.ylabel('Spending Score')





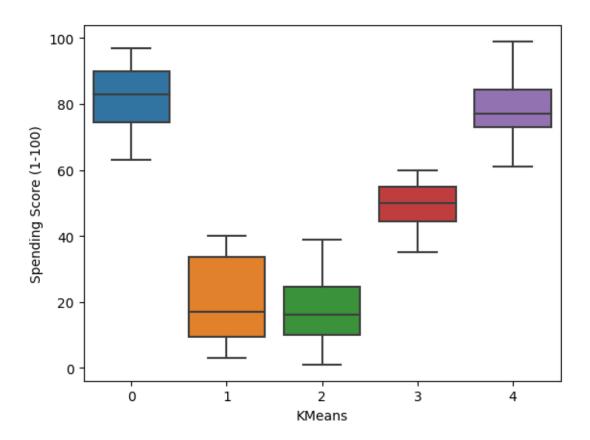
```
[29]: data['KMeans'].value_counts().plot.pie(autopct='%0.2f%%')
```

[29]: <Axes: ylabel='KMeans'>



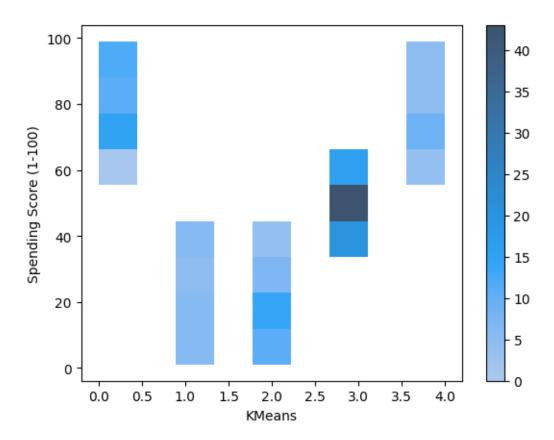
```
[30]: sns.boxplot(data=data,x='KMeans',y='Spending Score (1-100)')
```

[30]: <Axes: xlabel='KMeans', ylabel='Spending Score (1-100)'>



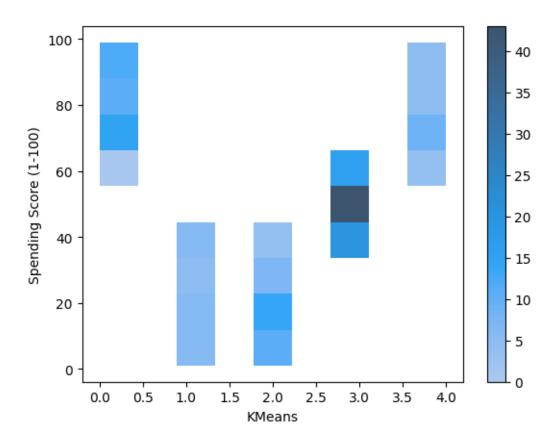
```
[32]: sns.histplot(data,x='KMeans',y='Spending Score (1-100)',cbar=True)
```

[32]: <Axes: xlabel='KMeans', ylabel='Spending Score (1-100)'>



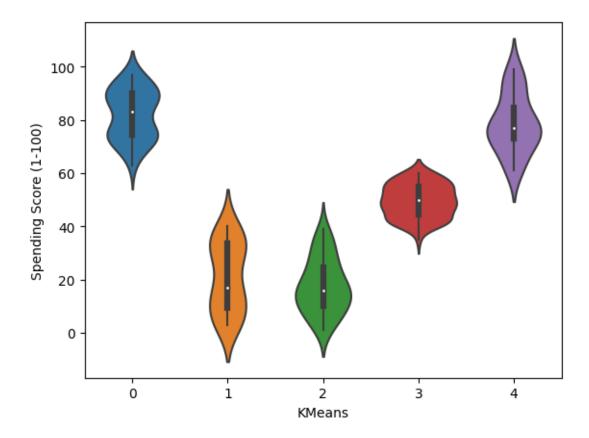
```
[33]: sns.histplot(data,x='KMeans',y='Spending Score (1-100)',cbar=True)
```

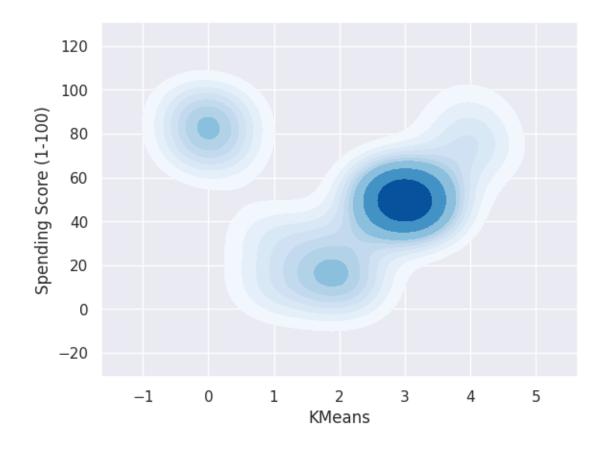
[33]: <Axes: xlabel='KMeans', ylabel='Spending Score (1-100)'>

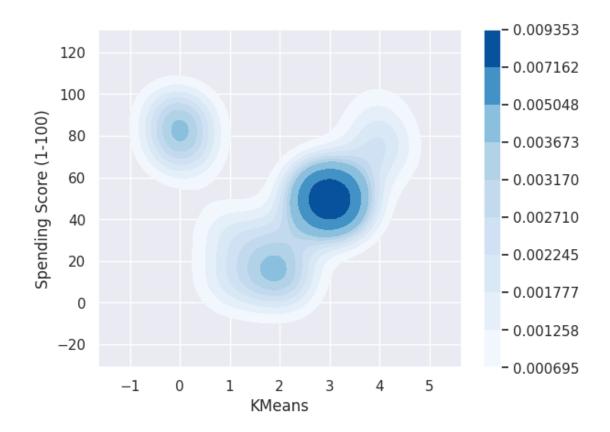


```
[34]: sns.violinplot(data=data, x="KMeans", y="Spending Score (1-100)") #to note the density
```

[34]: <Axes: xlabel='KMeans', ylabel='Spending Score (1-100)'>

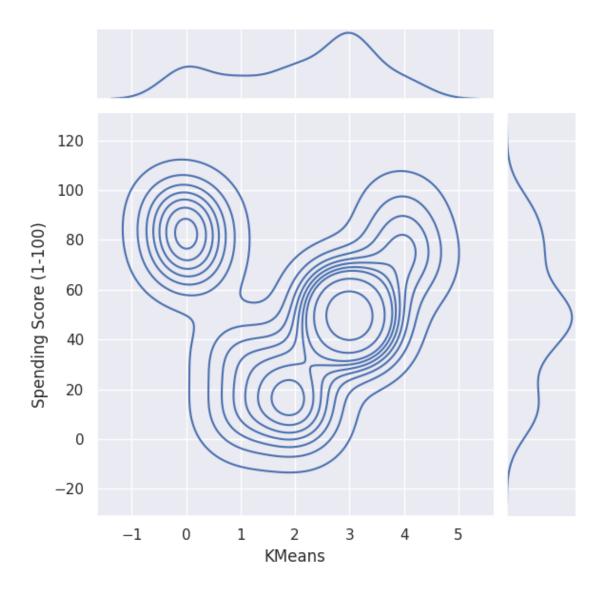






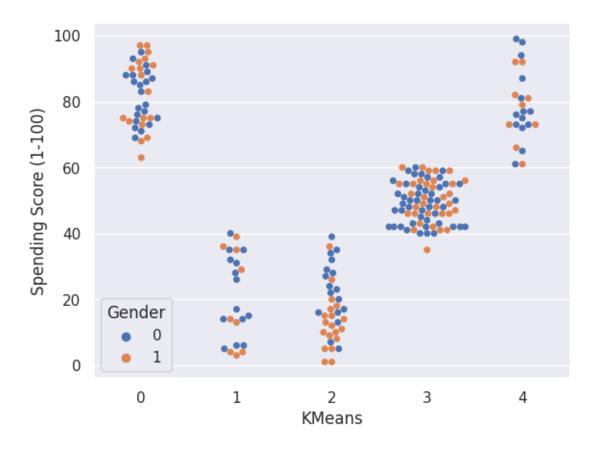
[38]: sns.jointplot(data=data,x="KMeans", y="Spending Score (1-100)",kind="kde")

[38]: <seaborn.axisgrid.JointGrid at 0x7d602fd1e3e0>



```
[40]: sns.swarmplot(data=data, x="KMeans", y="Spending Score (1-100)", hue='Gender')
```

[40]: <Axes: xlabel='KMeans', ylabel='Spending Score (1-100)'>



```
[41]: sns.catplot(data=data, x="KMeans", y="Spending Score (1-100)",⊔

⇒jitter=False,hue='Gender')
```

[41]: <seaborn.axisgrid.FacetGrid at 0x7d602e03f8e0>

