

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

import os
os.getcwd()

    '/content'

from sklearn.metrics import ConfusionMatrixDisplay, accuracy_score, classification_report
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive_bayes import MultinomialNB

df=pd.read_csv('/content/archive_mall.zip')

```

```
df.head()
```

| | CustomerID | Genre | 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 | |

```
df.shape
```

```
(200, 5)
```

```
x=df.iloc[:,3:]
```

```
x
```

| | Annual Income (k\$) | Spending Score (1-100) | |
|-----|---------------------|------------------------|--|
| 0 | 15 | 39 | |
| 1 | 15 | 81 | |
| 2 | 16 | 6 | |
| 3 | 16 | 77 | |
| 4 | 17 | 40 | |
| ... | ... | ... | |
| 195 | 120 | 79 | |
| 196 | 126 | 28 | |
| 197 | 126 | 74 | |
| 198 | 137 | 18 | |
| 199 | 137 | 83 | |

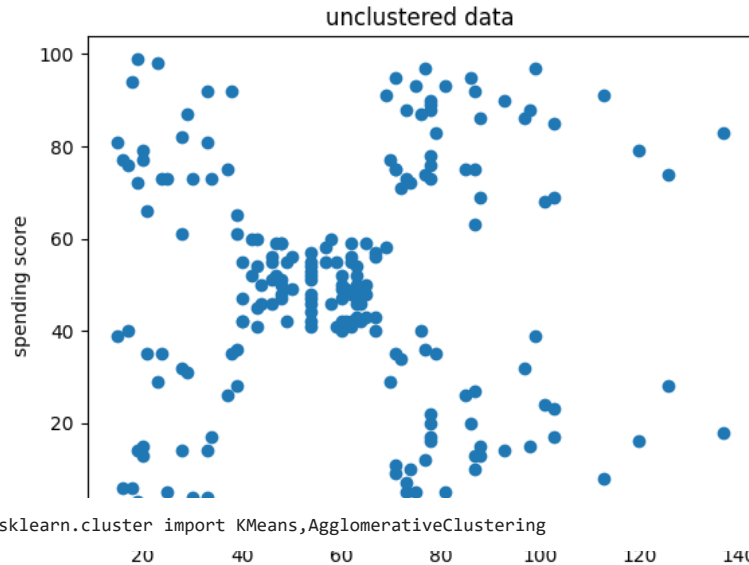
```
200 rows × 2 columns
```

```

plt.title('unclustered data')
plt.xlabel('annual income')
plt.ylabel('spending score')
plt.scatter(x['Annual Income (k$)'],x['Spending Score (1-100)'])

```

```
<matplotlib.collections.PathCollection at 0x7a45ad1ed960>
```



```
from sklearn.cluster import KMeans, AgglomerativeClustering
```

```
sse=[]
```

```
for k in range(1,16):
    km=KMeans(n_clusters=k)
    km.fit_predict(x)
    sse.append(km.inertia_)
```

[illegible]

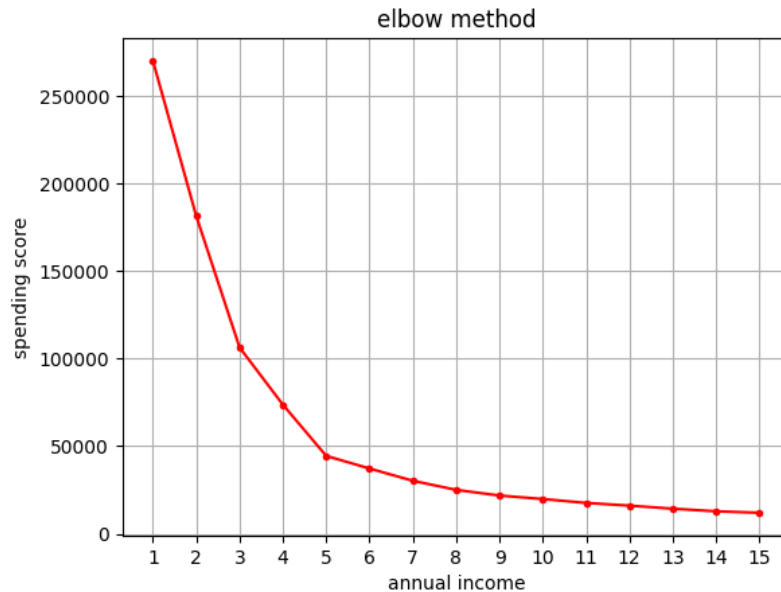
sse

[269981.28,
181363.59595959593,
106348.37306211122,
73679.78903948836,
44448.4554479337,
37233.814510710006,
30241.34361793658,
25043.970949607214,
21818.114588452176,
19829.031646581996,
17602.19046838677,
16049.714149902382,
14314.980233818465.

12816.898555873555,
11995.506623278681]

```
plt.title('elbow method')
plt.xlabel('annual income')
plt.ylabel('spending score')
plt.grid()
plt.xticks(range(1,16))
plt.plot(range(1,16),sse,marker='.',color='r')
```

```
[<matplotlib.lines.Line2D at 0x7a45acba4f40>]
```



```
silh=[]
```

```
from sklearn.metrics.cluster import silhouette_score
for k in range(2,16):
    km=KMeans(n_clusters=k)
    labels=km.fit_predict(x)
    score=silhouette_score(x,labels)
    silh.append(score)
```

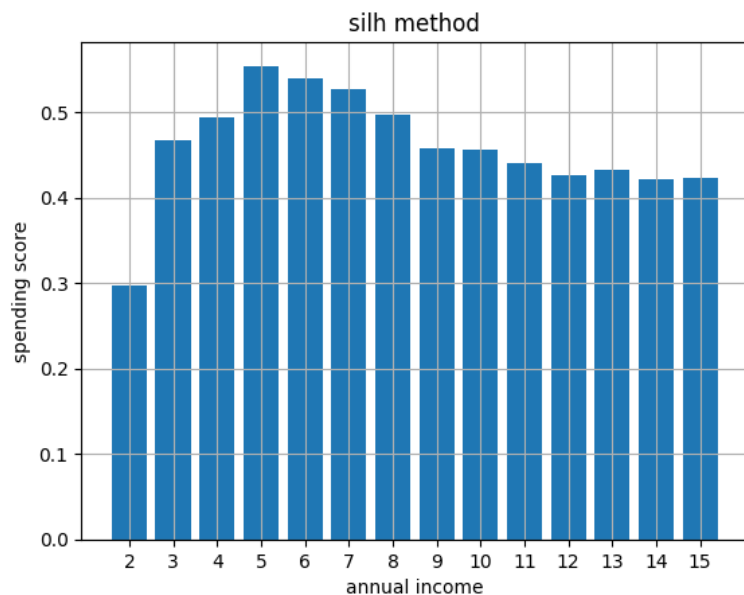
```
y:870: FutureWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress this warning.
```

silh

```
[0.2968969162503008,
0.46761358158775435,
0.4931963109249047,
0.553931997444648,
0.53976103063432,
0.5270287298101395,
0.4976260414857619,
0.45819645551960536,
0.45575852895571267,
0.4395970285642804,
0.4263917088390977,
0.43211339472104393,
0.42200427744178964,
0.4235485715125881]
```

```
plt.title('silh method')
plt.xlabel('annual income')
plt.ylabel('spending score')
plt.grid()
plt.xticks(range(2,16))
plt.bar(range(2,16),silh)
```

<BarContainer object of 14 artists>



```
km=KMeans(n_clusters=5)
```

```
labels=km.fit_predict(x)
```

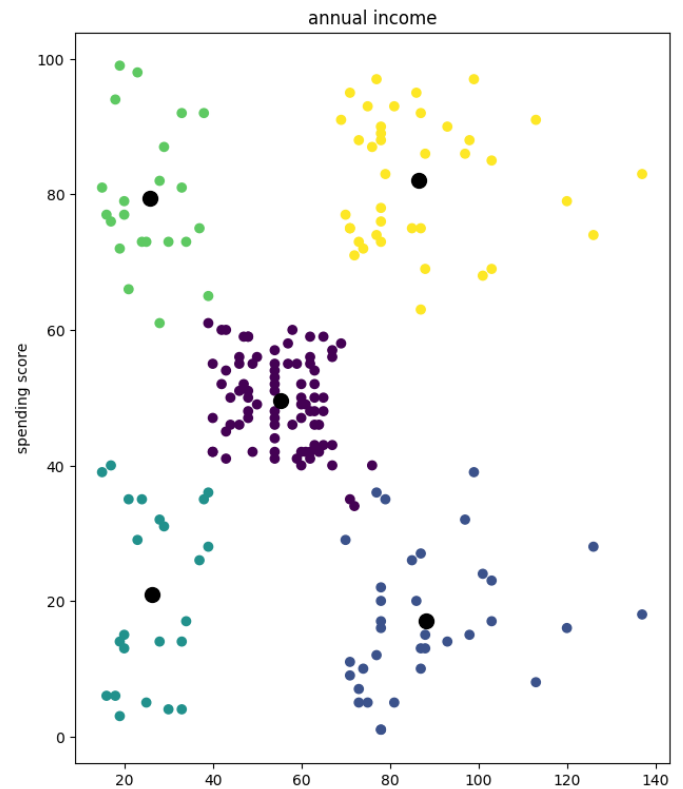
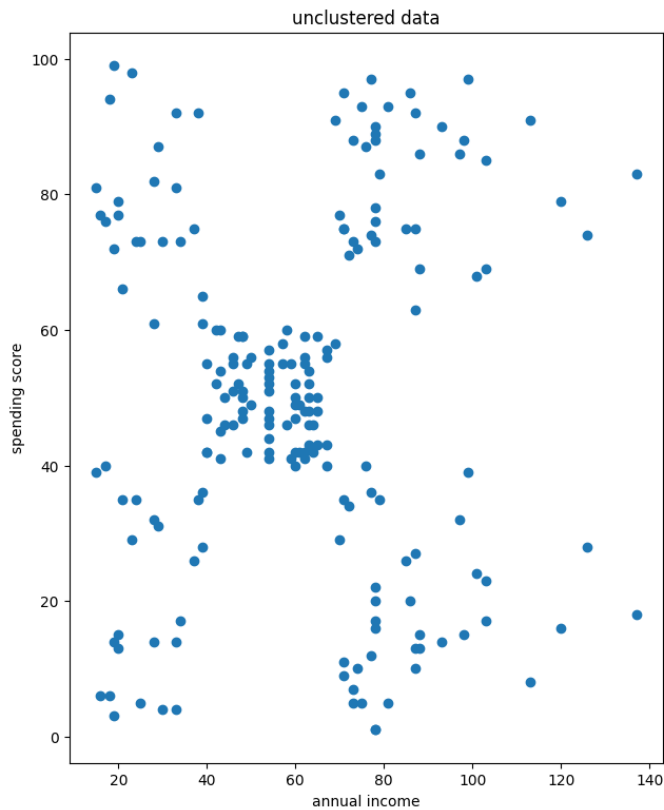
```
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from
warnings.warn(
```

```
cent=km.cluster_centers_
```

```
plt.figure(figsize=(16,9))
plt.subplot(1,2,1)
plt.title('unclustered data')
plt.xlabel('annual income')
plt.ylabel('spending score')
plt.scatter(x['Annual Income (k$)'],x['Spending Score (1-100)'])
```

```
plt.subplot(1,2,2)
plt.title('clustered data')
plt.title('annual income')
plt.ylabel('spending score')
plt.scatter(x['Annual Income (k$)'],x['Spending Score (1-100)'],c=labels)
plt.scatter(cent[:,0],cent[:,1],s=100,color='k')
```

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```
agl=AgglomerativeClustering(n_clusters=5)
```

```
alabels=agl.fit_predict(x)
```

```
plt.figure(figsize=(16,9))
plt.subplot(1,2,1)
plt.title('agl')
plt.xlabel('annual income')
plt.ylabel('spending score')
plt.scatter(x['Annual Income (k$)'],x['Spending Score (1-100)'],c=alabels)
```

```
plt.subplot(1,2,2)
plt.title('k means')
plt.title('annual income')
plt.ylabel('spending score')
plt.scatter(x['Annual Income (k$)'],x['Spending Score (1-100)'],c=labels)
plt.scatter(cent[:,0],cent[:,1],s=100,color='k')
```

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