

```
In [1]: import numpy as np
import pandas as pd
```

```
In [3]: df=pd.read_csv('student_clustering.csv')
```

```
In [4]: df.head()
```

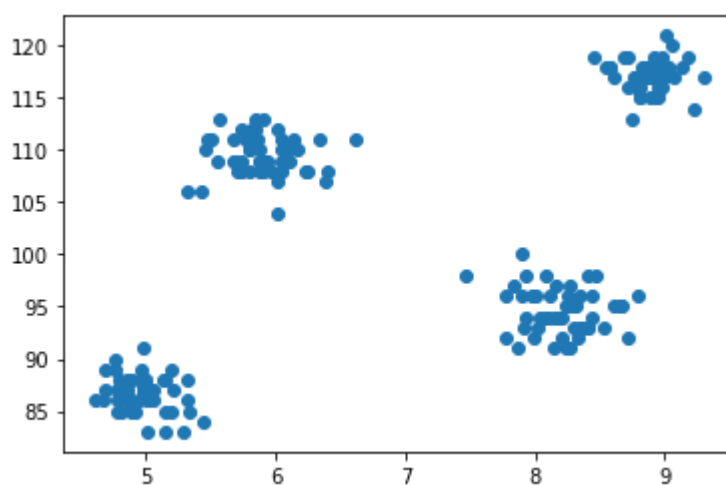
Out[4]:

	cgpa	iq
0	5.13	88
1	5.90	113
2	8.36	93
3	8.27	97
4	5.45	110

```
In [5]: import matplotlib.pyplot as plt
```

```
In [6]: plt.scatter(df['cgpa'],df['iq'])
```

Out[6]: <matplotlib.collections.PathCollection at 0x276c2060b20>



```
In [13]: from sklearn.cluster import KMeans
```

```
In [14]: wcss=[]
```

```
In [15]: for i in range(1,11):
    km=KMeans(n_clusters=i)
    km.fit_predict(df)
    wcss.append(km.inertia_)
```

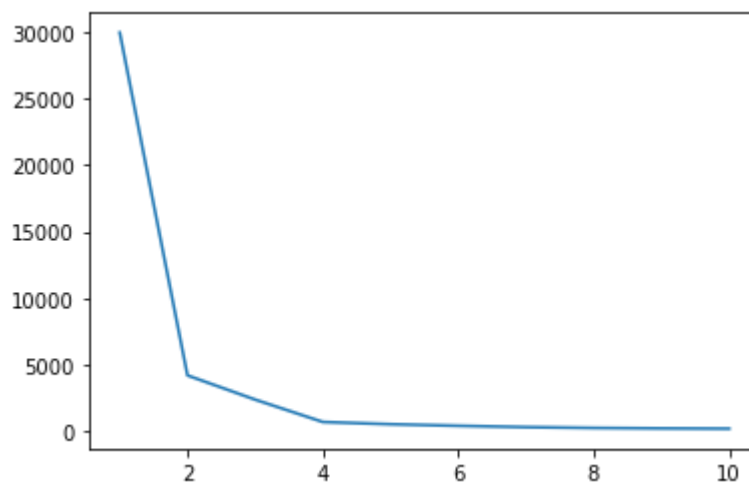
C:\Users\SHIVAM .A.R.GHORPADE\Anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:1036: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.
warnings.warn(

In [16]: wcss

Out[16]: [29957.898287999997,
4184.14127,
2364.005583420083,
681.96966,
514.1616803171114,
395.9605877691832,
295.4391895943192,
233.54082485509016,
199.9912003256784,
179.26308183060485]

In [17]: plt.plot(range(1,11),wcss)

Out[17]: [<matplotlib.lines.Line2D at 0x276c3fcf4c0>]



In [18]: x=df.iloc[:,:].values

In [24]: km=KMeans(n_clusters=4)
y_means=km.fit_predict(x)

In [26]: y_means

Out[26]: array([0, 1, 2, 2, 1, 1, 2, 3, 1, 2, 0, 1, 2, 0, 1, 2, 1, 2, 1, 1, 2, 0,
2, 0, 0, 2, 0, 3, 2, 1, 3, 1, 3, 1, 2, 2, 3, 1, 0, 1, 0, 2, 2, 0,
3, 3, 2, 1, 3, 1, 0, 0, 3, 2, 3, 1, 1, 3, 1, 3, 1, 2, 2, 3, 0, 3,
2, 0, 1, 2, 1, 3, 2, 0, 1, 3, 1, 3, 0, 2, 2, 3, 1, 0, 3, 0, 3, 1,
3, 1, 3, 3, 2, 0, 2, 2, 3, 2, 0, 3, 1, 0, 0, 3, 0, 0, 2, 0, 3, 3,
2, 3, 1, 1, 2, 3, 2, 1, 3, 0, 0, 1, 2, 3, 2, 0, 2, 1, 0, 2, 2, 1,
0, 0, 1, 3, 1, 0, 2, 2, 2, 0, 1, 0, 0, 3, 0, 3, 1, 0, 3, 0, 3, 3,
0, 2, 1, 3, 1, 2, 0, 3, 1, 2, 3, 0, 1, 0, 0, 3, 3, 1, 3, 0, 0, 2,
3, 1, 0, 3, 3, 1, 1, 1, 2, 0, 2, 2, 3, 1, 2, 2, 0, 0, 2, 0, 3, 1,
1, 3])

```
In [27]: x[y_means==0]
```

```
Out[27]: array([[ 5.13, 88. ],
 [ 4.6 , 86. ],
 [ 5.   , 88. ],
 [ 4.86, 86. ],
 [ 4.78, 87. ],
 [ 4.96, 88. ],
 [ 4.86, 87. ],
 [ 5.44, 84. ],
 [ 5.34, 85. ],
 [ 5.31, 86. ],
 [ 5.14, 83. ],
 [ 4.95, 86. ],
 [ 5.21, 87. ],
 [ 4.91, 85. ],
 [ 5.28, 83. ],
 [ 5.15, 88. ],
 [ 4.9 , 85. ],
 [ 4.89, 88. ],
 [ 5.05, 86. ],
 [ 4.98, 91. ],
 [ 5.01, 86. ],
 [ 4.95, 88. ],
 [ 4.96, 89. ],
 [ 4.85, 86. ],
 [ 4.76, 90. ],
 [ 4.98, 87. ],
 [ 4.78, 87. ],
 [ 5.2 , 85. ],
 [ 5.05, 87. ],
 [ 5.01, 83. ],
 [ 4.77, 86. ],
 [ 4.68, 87. ],
 [ 4.81, 85. ],
 [ 5.03, 87. ],
 [ 4.98, 87. ],
 [ 5.32, 88. ],
 [ 4.86, 88. ],
 [ 4.89, 85. ],
 [ 4.88, 86. ],
 [ 5.01, 86. ],
 [ 4.67, 86. ],
 [ 5.15, 85. ],
 [ 4.97, 88. ],
 [ 4.87, 88. ],
 [ 5.2 , 89. ],
 [ 4.99, 88. ],
 [ 4.79, 88. ],
 [ 4.76, 89. ],
 [ 4.78, 85. ],
 [ 4.68, 89. ]])
```

```
In [28]: x[y_means==1]
```

```
Out[28]: array([[ 5.9 , 113. ],
 [ 5.45, 110. ],
 [ 5.88, 109. ],
 [ 5.79, 110. ],
 [ 6.1 , 110. ],
 [ 5.71, 108. ],
 [ 5.5 , 111. ],
 [ 6.05, 111. ],
 [ 5.84, 113. ],
 [ 5.43, 106. ],
 [ 6.01, 112. ],
 [ 5.32, 106. ],
 [ 5.91, 108. ],
 [ 5.57, 113. ],
 [ 6.4 , 108. ],
 [ 5.67, 109. ],
 [ 6.05, 108. ],
 [ 5.85, 111. ],
 [ 5.87, 109. ],
 [ 6.02, 104. ],
 [ 5.77, 111. ],
 [ 6.06, 109. ],
 [ 5.55, 109. ],
 [ 5.81, 112. ],
 [ 5.47, 111. ],
 [ 5.74, 109. ],
 [ 5.8 , 108. ],
 [ 5.88, 110. ],
 [ 5.91, 109. ],
 [ 5.67, 111. ],
 [ 5.74, 108. ],
 [ 5.69, 109. ],
 [ 6.05, 109. ],
 [ 6.14, 111. ],
 [ 5.74, 112. ],
 [ 5.94, 109. ],
 [ 5.86, 111. ],
 [ 6.38, 107. ],
 [ 6.61, 111. ],
 [ 6.04, 110. ],
 [ 6.24, 108. ],
 [ 6.1 , 109. ],
 [ 5.8 , 110. ],
 [ 5.87, 108. ],
 [ 5.97, 108. ],
 [ 6.17, 110. ],
 [ 6.01, 107. ],
 [ 6.33, 111. ],
 [ 5.85, 112. ],
 [ 6.23, 108. ]])
```

```
In [29]: x[y_means==2]
```

```
Out[29]: array([[ 8.36,  93. ],
 [ 8.27,  97. ],
 [ 8.41,  98. ],
 [ 8.09,  94. ],
 [ 8.16,  97. ],
 [ 8.31,  95. ],
 [ 7.87,  91. ],
 [ 7.47,  98. ],
 [ 7.78,  92. ],
 [ 7.93,  98. ],
 [ 8.04,  94. ],
 [ 7.77,  96. ],
 [ 8.   ,  96. ],
 [ 8.43,  96. ],
 [ 8.02,  93. ],
 [ 8.14,  94. ],
 [ 8.12,  96. ],
 [ 8.34,  96. ],
 [ 8.65,  95. ],
 [ 8.53,  93. ],
 [ 8.29,  95. ],
 [ 7.93,  94. ],
 [ 8.72,  92. ],
 [ 8.14,  91. ],
 [ 8.2   ,  92. ],
 [ 8.67,  95. ],
 [ 8.18,  94. ],
 [ 8.61,  95. ],
 [ 7.99,  92. ],
 [ 8.08,  94. ],
 [ 8.26,  91. ],
 [ 8.25,  95. ],
 [ 8.4   ,  93. ],
 [ 7.84,  97. ],
 [ 8.08,  98. ],
 [ 8.25,  96. ],
 [ 8.3   ,  93. ],
 [ 7.9   , 100. ],
 [ 7.97,  96. ],
 [ 8.21,  94. ],
 [ 8.23,  95. ],
 [ 8.35,  93. ],
 [ 8.33,  92. ],
 [ 8.46,  98. ],
 [ 7.89,  96. ],
 [ 7.91,  93. ],
 [ 8.23,  91. ],
 [ 8.4   ,  93. ],
 [ 8.44,  94. ],
 [ 8.79,  96. ]])
```

```
In [30]: x[y_means==3]
```

```
Out[30]: array([[ 8.8 , 115. ],
 [ 9.18, 119. ],
 [ 8.86, 117. ],
 [ 8.83, 118. ],
 [ 8.56, 118. ],
 [ 8.96, 116. ],
 [ 8.78, 116. ],
 [ 8.45, 119. ],
 [ 8.79, 116. ],
 [ 8.81, 115. ],
 [ 8.88, 115. ],
 [ 9.07, 117. ],
 [ 8.92, 118. ],
 [ 8.75, 113. ],
 [ 8.71, 116. ],
 [ 8.86, 118. ],
 [ 9.3 , 117. ],
 [ 9.01, 121. ],
 [ 8.97, 116. ],
 [ 9. , 117. ],
 [ 8.76, 117. ],
 [ 8.78, 117. ],
 [ 9.23, 114. ],
 [ 9.03, 118. ],
 [ 9.13, 118. ],
 [ 8.91, 119. ],
 [ 8.98, 118. ],
 [ 9.03, 118. ],
 [ 8.86, 117. ],
 [ 8.89, 118. ],
 [ 8.97, 117. ],
 [ 8.72, 119. ],
 [ 8.93, 118. ],
 [ 8.58, 118. ],
 [ 8.94, 117. ],
 [ 8.6 , 117. ],
 [ 8.77, 117. ],
 [ 8.81, 116. ],
 [ 8.54, 118. ],
 [ 8.97, 119. ],
 [ 8.91, 117. ],
 [ 8.68, 119. ],
 [ 9.06, 120. ],
 [ 8.9 , 117. ],
 [ 8.94, 115. ],
 [ 8.91, 115. ],
 [ 8.91, 117. ],
 [ 8.95, 116. ],
 [ 8.57, 118. ],
 [ 8.82, 117. ]])
```

```
In [31]: x[y_means==3,0]
```

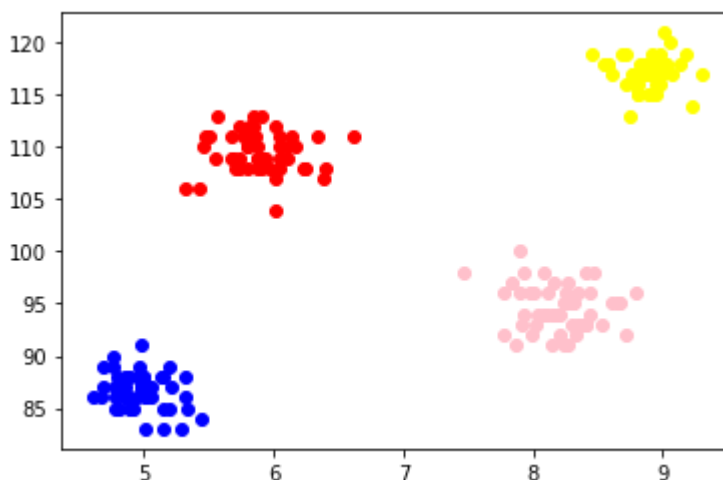
```
Out[31]: array([8.8 , 9.18, 8.86, 8.83, 8.56, 8.96, 8.78, 8.45, 8.79, 8.81, 8.88,  
                9.07, 8.92, 8.75, 8.71, 8.86, 9.3 , 9.01, 8.97, 9. , 8.76, 8.78,  
                9.23, 9.03, 9.13, 8.91, 8.98, 9.03, 8.86, 8.89, 8.97, 8.72, 8.93,  
                8.58, 8.94, 8.6 , 8.77, 8.81, 8.54, 8.97, 8.91, 8.68, 9.06, 8.9 ,  
                8.94, 8.91, 8.91, 8.95, 8.57, 8.82])
```

```
In [32]: x[y_means==3,1]
```

```
Out[32]: array([115., 119., 117., 118., 118., 116., 116., 119., 116., 115., 115.,  
                117., 118., 113., 116., 118., 117., 121., 116., 117., 117., 117.,  
                114., 118., 118., 119., 118., 118., 117., 118., 117., 119., 118.,  
                118., 117., 117., 117., 116., 118., 119., 117., 119., 120., 117.,  
                115., 115., 117., 116., 118., 117.])
```

```
In [39]: plt.scatter(x[y_means==0,0],x[y_means==0,1],color='blue')  
plt.scatter(x[y_means==1,0],x[y_means==1,1],color='red')  
plt.scatter(x[y_means==2,0],x[y_means==2,1],color='pink')  
plt.scatter(x[y_means==3,0],x[y_means==3,1],color='yellow')
```

```
Out[39]: <matplotlib.collections.PathCollection at 0x276c7b2da60>
```



```
In [ ]:
```