Breast Cancer Prediction

In [1]: import pandas as pd
 from matplotlib import pyplot as plt
 %matplotlib inline

Out[2]:

	Id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness
0	842302	М	17.99	10.38	122.80	1001.0	(
1	842517	М	20.57	17.77	132.90	1326.0	(
2	84300903	М	19.69	21.25	130.00	1203.0	(
3	84348301	М	11.42	20.38	77.58	386.1	(
4	84358402	М	20.29	14.34	135.10	1297.0	(
564	926424	М	21.56	22.39	142.00	1479.0	
565	926682	М	20.13	28.25	131.20	1261.0	(
566	926954	М	16.60	28.08	108.30	858.1	(
567	927241	М	20.60	29.33	140.10	1265.0	(
568	92751	В	7.76	24.54	47.92	181.0	(

569 rows × 33 columns

→

In [3]: df.head()

Out[3]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
0	842302	М	17.99	10.38	122.80	1001.0	0.1
1	842517	М	20.57	17.77	132.90	1326.0	0.0
2	84300903	М	19.69	21.25	130.00	1203.0	0.1
3	84348301	М	11.42	20.38	77.58	386.1	0.1
4	84358402	М	20.29	14.34	135.10	1297.0	0.1

5 rows × 33 columns

In [4]: df.tail()

Out[4]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
564	926424	М	21.56	22.39	142.00	1479.0	0.1
565	926682	М	20.13	28.25	131.20	1261.0	0.0
566	926954	М	16.60	28.08	108.30	858.1	0.0
567	927241	М	20.60	29.33	140.10	1265.0	0.1
568	92751	В	7.76	24.54	47.92	181.0	0.0

5 rows × 33 columns

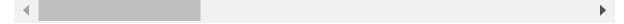
→

In [5]: df.drop(['Unnamed: 32'],axis=1)

Out[5]:

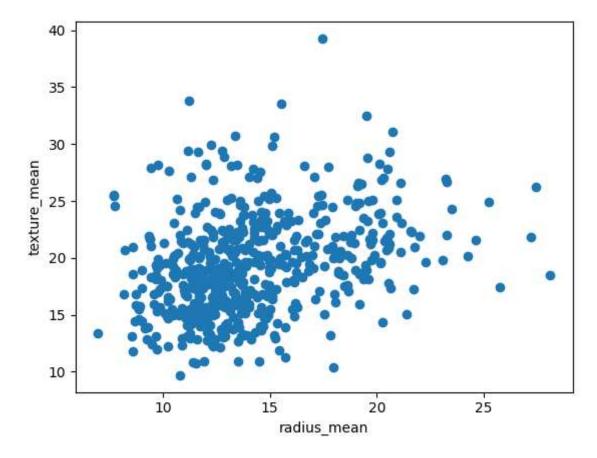
id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness
842302	М	17.99	10.38	122.80	1001.0	(
842517	М	20.57	17.77	132.90	1326.0	(
84300903	М	19.69	21.25	130.00	1203.0	(
84348301	М	11.42	20.38	77.58	386.1	(
84358402	М	20.29	14.34	135.10	1297.0	(
926424	М	21.56	22.39	142.00	1479.0	
926682	М	20.13	28.25	131.20	1261.0	(
926954	М	16.60	28.08	108.30	858.1	(
927241	М	20.60	29.33	140.10	1265.0	(
92751	В	7.76	24.54	47.92	181.0	(
	842302 842517 84300903 84348301 84358402 926424 926682 926954 927241	842302 M 842517 M 84300903 M 84348301 M 84358402 M 926424 M 926682 M 926954 M	842302 M 17.99 842517 M 20.57 84300903 M 19.69 84348301 M 11.42 84358402 M 20.29 926424 M 21.56 926682 M 20.13 926954 M 16.60 927241 M 20.60	842302 M 17.99 10.38 842517 M 20.57 17.77 84300903 M 19.69 21.25 84348301 M 11.42 20.38 84358402 M 20.29 14.34 926424 M 21.56 22.39 926682 M 20.13 28.25 926954 M 16.60 28.08 927241 M 20.60 29.33	842302 M 17.99 10.38 122.80 842517 M 20.57 17.77 132.90 84300903 M 19.69 21.25 130.00 84348301 M 11.42 20.38 77.58 84358402 M 20.29 14.34 135.10 926424 M 21.56 22.39 142.00 926682 M 20.13 28.25 131.20 926954 M 16.60 28.08 108.30 927241 M 20.60 29.33 140.10	842302 M 17.99 10.38 122.80 1001.0 842517 M 20.57 17.77 132.90 1326.0 84300903 M 19.69 21.25 130.00 1203.0 84348301 M 11.42 20.38 77.58 386.1 84358402 M 20.29 14.34 135.10 1297.0 926424 M 21.56 22.39 142.00 1479.0 926682 M 20.13 28.25 131.20 1261.0 926954 M 16.60 28.08 108.30 858.1 927241 M 20.60 29.33 140.10 1265.0

569 rows × 32 columns



```
In [6]: plt.scatter(df["radius_mean"],df["texture_mean"])
    plt.xlabel("radius_mean")
    plt.ylabel("texture_mean")
```

Out[6]: Text(0, 0.5, 'texture_mean')



```
In [7]: from sklearn.cluster import KMeans
    km=KMeans()
    km
```

Out[7]:

▼ KMeans

KMeans()

```
In [8]: y_predicted=km.fit_predict(df[["radius_mean","texture_mean"]])
y_predicted
```

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earn\cluster_kmeans.py:870: FutureWarning: The default value of `n_init` wi
ll change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to
suppress the warning
warnings.warn(

```
Out[8]: array([1, 4, 4, 7, 4, 1, 4, 3, 0, 0, 3, 3, 2, 3, 0, 5, 3, 3, 4, 1, 1, 6,
               1, 2, 3, 1, 3, 4, 0, 1, 2, 7, 3, 2, 3, 3, 3, 7, 0, 3, 0, 0, 2, 3,
               0, 4, 7, 7, 6, 0, 0, 1, 7, 4, 3, 7, 4, 3, 7, 6, 6, 7, 0, 6, 0, 0,
               7, 7, 7, 1, 4, 6, 2, 1, 7, 3, 6, 1, 2, 7, 0, 1, 2, 2, 6, 4, 3, 2,
               0, 1, 0, 3, 1, 7, 3, 2, 7, 7, 6, 3, 0, 6, 7, 7, 7, 1, 7, 7, 4, 0,
               7, 0, 3, 7, 6, 0, 6, 1, 3, 3, 6, 4, 4, 1, 1, 1, 0, 4, 1, 2, 6, 3,
               3, 1, 4, 0, 7, 6, 1, 6, 6, 3, 7, 1, 6, 6, 7, 3, 1, 7, 0, 7, 6, 6,
               1, 7, 3, 3, 6, 6, 7, 4, 4, 0, 4, 3, 6, 3, 2, 1, 6, 7, 1, 6, 6, 6,
               7, 3, 0, 6, 4, 2, 3, 6, 3, 6, 4, 7, 7, 1, 0, 0, 7, 5, 0, 1, 0, 3,
               4, 3, 7, 3, 2, 0, 7, 1, 7, 3, 0, 1, 4, 7, 4, 2, 0, 1, 7, 7, 4, 2,
               1, 1, 7, 3, 1, 1, 6, 1, 0, 0, 3, 5, 5, 2, 6, 3, 2, 4, 5, 5, 1, 6,
               7, 0, 2, 7, 7, 6, 0, 6, 2, 7, 4, 1, 4, 1, 2, 1, 3, 5, 2, 3, 3, 3,
               3, 2, 7, 0, 1, 7, 1, 6, 4, 6, 2, 7, 6, 4, 7, 1, 2, 6, 4, 3, 1, 7,
               0, 6, 7, 7, 3, 3, 1, 7, 6, 1, 6, 7, 7, 0, 4, 7, 2, 7, 7, 0, 1, 6,
               6, 6, 7, 1, 6, 6, 7, 7, 6, 4, 7, 7, 6, 4, 6, 4, 6, 7, 1, 7,
               1, 7, 7, 6, 7, 3, 1, 4, 7, 2, 1, 7, 6, 4, 6, 6, 7, 1, 6, 6, 7, 3,
               4, 0, 6, 7, 7, 1, 6, 7, 7, 0, 7, 3, 1, 4, 2, 7, 4, 4, 3, 1, 4, 4,
               1, 1, 7, 5, 1, 7, 6, 6, 0, 7, 1, 0, 6, 1, 6, 2, 6, 7, 3, 4,
               7, 7, 6, 7, 3, 6, 7, 1, 6, 7, 1, 0, 3, 7, 7, 7, 0, 3, 5, 0, 0, 3,
               6, 0, 7, 1, 6, 7, 7, 0, 6, 0, 7, 7, 3, 7, 4, 4, 1, 3, 7, 1, 3, 1,
               7, 2, 1, 7, 4, 0, 2, 1, 3, 4, 0, 2, 5, 1, 7, 5, 5, 0, 0, 5, 2, 2,
               5, 7, 7, 7, 0, 7, 3, 7, 7, 5, 1, 5, 6, 1, 3, 1, 6, 3, 7, 3, 1, 7,
               1, 7, 1, 4, 7, 3, 0, 1, 3, 6, 0, 3, 7, 7, 4, 4, 1, 0, 1, 4, 6, 6,
               7, 7, 1, 0, 6, 1, 3, 1, 3, 7, 4, 4, 7, 7, 6, 4, 7, 7, 6, 6, 7, 6,
               1, 6, 7, 7, 1, 4, 7, 4, 0, 0, 0, 6, 0, 0, 5, 3, 0, 7, 7, 7, 0,
               0, 0, 5, 0, 5, 5, 7, 5, 0, 0, 5, 5, 5, 2, 4, 2, 5, 2, 0])
```

```
In [9]: df["cluster"]=y_predicted
df.head()
```

Out[9]:

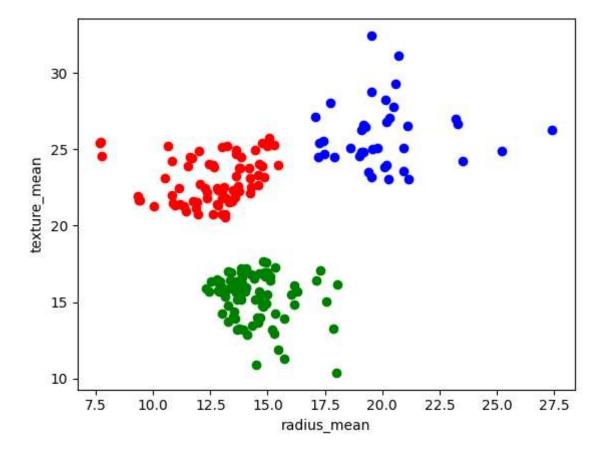
	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
0	842302	М	17.99	10.38	122.80	1001.0	0.1
1	842517	М	20.57	17.77	132.90	1326.0	0.0
2	84300903	М	19.69	21.25	130.00	1203.0	0.1
3	84348301	М	11.42	20.38	77.58	386.1	0.1
4	84358402	М	20.29	14.34	135.10	1297.0	0.1

5 rows × 34 columns

→

```
In [10]: df1=df[df.cluster==0]
    df2=df[df.cluster==1]
    df3=df[df.cluster==2]
    plt.scatter(df1["radius_mean"],df1["texture_mean"],color="red")
    plt.scatter(df2["radius_mean"],df2["texture_mean"],color="green")
    plt.scatter(df3["radius_mean"],df3["texture_mean"],color="blue")
    plt.xlabel("radius_mean")
    plt.ylabel("texture_mean")
```

Out[10]: Text(0, 0.5, 'texture_mean')



```
In [11]: from sklearn.preprocessing import MinMaxScaler
    scaler=MinMaxScaler()
    scaler.fit(df[["texture_mean"]])
    df["texture_mean"]=scaler.transform(df[["texture_mean"]])
    df.head()
```

Out[11]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
0	842302	М	17.99	0.022658	122.80	1001.0	0.1
1	842517	М	20.57	0.272574	132.90	1326.0	0.0
2	84300903	М	19.69	0.390260	130.00	1203.0	0.1
3	84348301	М	11.42	0.360839	77.58	386.1	0.1
4	84358402	М	20.29	0.156578	135.10	1297.0	0.1

5 rows × 34 columns

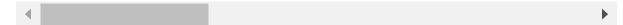
◆

```
In [12]: scaler.fit(df[["radius_mean"]])
    df["radius_mean"]=scaler.transform(df[["radius_mean"]])
    df.head()
```

Out[12]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
0	842302	М	0.521037	0.022658	122.80	1001.0	0.1
1	842517	М	0.643144	0.272574	132.90	1326.0	0.0
2	84300903	М	0.601496	0.390260	130.00	1203.0	0.1
3	84348301	М	0.210090	0.360839	77.58	386.1	0.1
4	84358402	М	0.629893	0.156578	135.10	1297.0	0.1

5 rows × 34 columns



```
In [13]: y_predicted=km.fit_predict(df[["radius_mean","texture_mean"]])
y_predicted
```

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earn\cluster_kmeans.py:870: FutureWarning: The default value of `n_init` wi
ll change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to
suppress the warning
warnings.warn(

Out[13]: array([4, 0, 0, 1, 0, 4, 0, 6, 6, 6, 6, 4, 7, 6, 6, 2, 6, 6, 0, 4, 4, 3, 4, 5, 0, 0, 6, 0, 6, 0, 7, 1, 7, 7, 0, 0, 6, 1, 6, 6, 6, 1, 7, 6, 6, 0, 3, 1, 3, 6, 1, 4, 1, 0, 6, 1, 0, 6, 1, 3, 3, 1, 6, 3, 6, 6, 1, 1, 3, 4, 0, 3, 7, 4, 4, 0, 4, 0, 7, 1, 1, 4, 5, 7, 3, 0, 6, 7, 6, 4, 6, 6, 4, 1, 6, 7, 1, 1, 3, 6, 6, 3, 1, 1, 1, 4, 1, 1, 5, 1, 3, 1, 6, 1, 3, 1, 3, 4, 6, 0, 3, 0, 5, 4, 4, 4, 6, 0, 4, 7, 3, 6, 6, 4, 0, 6, 1, 3, 4, 3, 3, 0, 1, 4, 3, 3, 1, 6, 4, 4, 6, 1, 3, 3, 4, 1, 0, 0, 3, 3, 1, 0, 0, 6, 5, 6, 3, 0, 7, 4, 3, 6, 4, 3, 3, 3, 1, 0, 6, 4, 5, 7, 6, 3, 6, 3, 0, 1, 1, 4, 6, 6, 1, 2, 6, 4, 6, 0, 7, 6, 1, 0, 5, 6, 1, 4, 1, 0, 6, 4, 5, 1, 5, 7, 6, 4, 1, 1, 0, 7, 4, 4, 1, 6, 4, 4, 3, 4, 6, 6, 0, 2, 2, 7, 3, 6, 5, 5, 2, 2, 4, 4, 1, 6, 7, 1, 4, 4, 2, 3, 5, 1, 0, 0, 0, 4, 7, 4, 6, 2, 7, 7, 0, 6, 0, 7, 1, 6, 4, 1, 4, 3, 5, 3, 7, 1, 3, 0, 4, 4, 7, 3, 0, 0, 4, 1, 1, 4, 1, 1, 6, 6, 4, 1, 4, 4, 3, 1, 4, 1, 0, 1, 7, 1, 1, 2, 4, 4, 1, 4, 4, 3, 1, 1, 3, 0, 1, 1, 3, 0, 4, 5, 3, 1, 4, 1, 6, 6, 4, 1, 1, 3, 1, 0, 4, 0, 1, 5, 4, 3, 3, 0, 3, 3, 1, 4, 3, 3, 1, 6, 5, 6, 3, 1, 1, 4, 3, 1, 1, 6, 1, 0, 4, 5, 7, 1, 5, 5, 6, 4, 0, 0, 4, 4, 1, 2, 4, 1, 3, 3, 6, 1, 4, 6, 3, 4, 3, 7, 3, 3, 6, 5, 1, 4, 1, 3, 1, 0, 3, 1, 4, 4, 1, 4, 6, 0, 1, 1, 1, 1, 6, 2, 1, 4, 1, 1, 4, 3, 6, 1, 1, 3, 1, 3, 1, 6, 1, 0, 0, 4, 6, 1, 4, 6, 4, 1, 7, 4, 1, 0, 2, 7, 4, 6, 5, 1, 7, 2, 4, 1, 2, 2, 2, 2, 2, 7, 5, 2, 1, 1, 6, 6, 1, 7, 1, 1, 2, 4, 2, 3, 4, 6, 4, 3, 0, 1, 6, 4, 4, 4, 4, 4, 0, 3, 0, 6, 4, 0, 3, 6, 6, 1, 1, 0, 5, 4, 6, 4, 5, 3, 3, 1, 1, 4, 6, 3, 4, 6, 4, 6, 1, 0, 0, 1, 4, 3, 5, 1, 1, 3, 3, 1, 3,

```
In [14]: df["New Cluster"]=y_predicted
df.head()
```

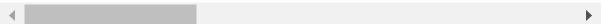
1, 1, 2, 1, 2, 2, 1, 2, 6, 2, 2, 2, 2, 7, 5, 7, 7, 7, 2])

Out[14]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
0	842302	М	0.521037	0.022658	122.80	1001.0	0.1
1	842517	М	0.643144	0.272574	132.90	1326.0	0.0
2	84300903	М	0.601496	0.390260	130.00	1203.0	0.1
3	84348301	М	0.210090	0.360839	77.58	386.1	0.1
4	84358402	М	0.629893	0.156578	135.10	1297.0	0.1

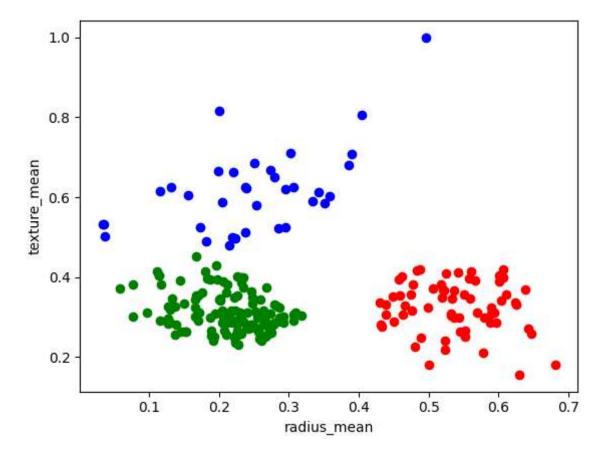
4, 3, 1, 1, 4, 0, 1, 5, 6, 2, 2, 2, 3, 6, 6, 2, 6, 6, 3, 3, 1, 2,

5 rows × 35 columns

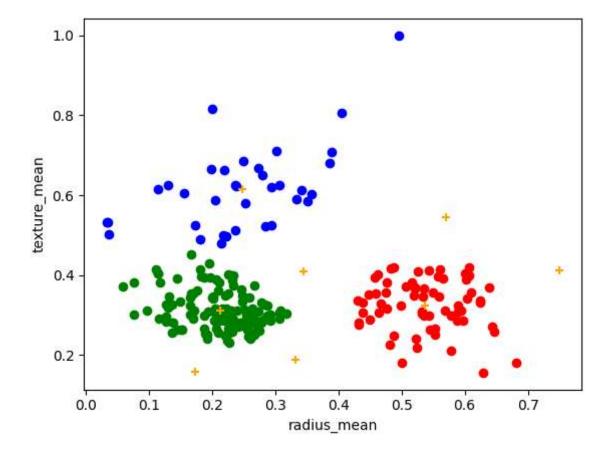


```
In [15]: df1=df[df["New Cluster"]==0]
    df2=df[df["New Cluster"]==1]
    df3=df[df["New Cluster"]==2]
    plt.scatter(df1["radius_mean"],df1["texture_mean"],color="red")
    plt.scatter(df2["radius_mean"],df2["texture_mean"],color="green")
    plt.scatter(df3["radius_mean"],df3["texture_mean"],color="blue")
    plt.xlabel("radius_mean")
    plt.ylabel("texture_mean")
```

Out[15]: Text(0, 0.5, 'texture_mean')



Out[17]: Text(0, 0.5, 'texture_mean')



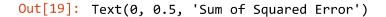
```
In [18]: k_rng=range(1,10)
sse=[]
```

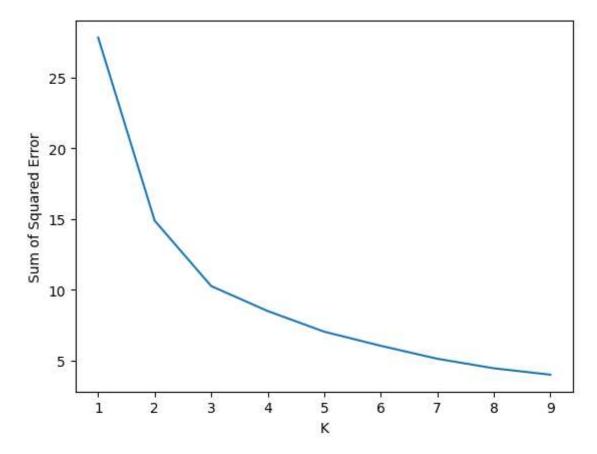
```
In [19]: | for k in k rng:
          km=KMeans(n clusters=k)
          km.fit(df[["radius_mean","texture mean"]])
          sse.append(km.inertia )
         #km.inertia will give you the value of sum of square error
         print(sse)
         plt.plot(k rng,sse)
         plt.xlabel("K")
         plt.ylabel("Sum of Squared Error")
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to
         suppress the warning
           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n init` explicitly to
         suppress the warning
           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n init` explicitly to
         suppress the warning
           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\ kmeans.py:870: FutureWarning: The default value of `n init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n init` explicitly to
         suppress the warning
           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\ kmeans.py:870: FutureWarning: The default value of `n init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n init` explicitly to
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           warnings.warn(
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         earn\cluster\ kmeans.py:870: FutureWarning: The default value of `n init` wi
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         suppress the warning
           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n init` explicitly to
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           warnings.warn(
         C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
         earn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` wi
         ll change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to
         suppress the warning
           warnings.warn(
         [27.81750759504308, 14.87203295827117, 10.25325555898541, 8.489103553996125,
```

7.030202097311372, 6.032638028620887, 5.117927753802227, 4.444537230930339,

3.9916276477713906]

C:\Users\venka\AppData\Local\Programs\Python\Python311\Lib\site-packages\skl
earn\cluster_kmeans.py:870: FutureWarning: The default value of `n_init` wi
ll change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to
suppress the warning
warnings.warn(





CONCLUSION

For the given dataset we can use multiple models, for that models we get different types of accuracies but that accuracies is not good so, that's why we will take it as a clustering and done with K-Means Clustering

In []:	