Problem Statement:

To Predict How Best the Datafits and To predict the BreastCancer based on the given Features

Importing All The Required Libraries

In [1]:

import pandas as pd
from matplotlib import pyplot as plt
%matplotlib inline

STEP-1: Data collection

In [2]:

df=pd.read_csv(r"C:\Users\thara\Downloads\BreastCancerPrediction.csv")
df

Out[2]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothn	
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								

STEP-2:Data Collection And Preprocessing

In [3]:

df.head()

Out[3]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothnes
0	842302	М	17.99	10.38	122.80	1001.0	
1	842517	M	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	

5 rows × 33 columns

→

In [4]:

df.tail()

Out[4]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothnes
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	M	20.60	29.33	140.10	1265.0	
568	92751	В	7.76	24.54	47.92	181.0	

5 rows × 33 columns

→

In [5]:

df.shape

Out[5]:

(569, 33)

In [6]:

df.describe

Out[6]:

	nd method NDF			id diagnosi	s radius_mean te
0	e_mean perin 842302	meter_mean M	area_mean 17.99	10.38	122.80 100
1.0	\ 842517	М	20.57	17.77	132.90 132
6.0	84300903	М	19.69	21.25	130.00 120
3.0	84348301	М	11.42	20.38	77.58 38
6.1 4 7.0	84358402	М	20.29	14.34	135.10 129
• •	•••	•••	•••	•••	
564	926424	М	21.56	22.39	142.00 147
9.0 565	926682	М	20.13	28.25	131.20 126
1.0 566	926954	М	16.60	28.08	108.30 85
8.1 567	927241	М	20.60	29.33	140.10 126
5.0 568 1.0	92751	В	7.76	24.54	47.92 18
1.0	smoothnoss n	moan compa	actnoss moon	concavity moan	concave points_mea
n	_	1840	0.27760	0.30010	0.1471
0 0 \ 1					
7 2		8474	0.07864	0.08690	0.0701
0	0.16		0.15990	0.19740	0.1279
3 0	0.14		0.28390	0.24140	0.1052
4 0	0.16	3030	0.13280	0.19800	0.1043
• •		•••	•••	•••	
564 0	0.11	1100	0.11590	0.24390	0.1389
565 1	0.09	9780	0.10340	0.14400	0.0979
566 2	0.08	3455	0.10230	0.09251	0.0530
567 0	0.11	L780	0.27700	0.35140	0.1520
568 0	0.05	5263	0.04362	0.00000	0.0000
	texture	e_worst pe	erimeter_worst	area_worst sm	noothness_worst
0 1	•••	17.33 23.41	184.60 158.80		0.16220 \ 0.12380
2	•••	25.53	152.50	1709.0	0.14440
3	• • •	26.50	98.87	567.7	0.20980
4	•••	16.67	152.20	1575.0	0.13740
 564	• • •	26.40	166.10	 2027.0	0.14100
565	• • •	38.25	155.00	1731.0	0.11660
566	• • •	34.12	126.70	1124.0	0.11390
567	• • •	39.42	184.60	1821.0	0.16500

30.37

59.16

268.6

0.08996

df.:	/]: .compactness_worst conca info()	vicy_worse cone	ave point	.5_WOI 3 C	Symmetry_wo
§}a:	0.66560 ss 'pandas.core.frame.Dat	0.7119 aFrame'>		0.2654	0.4
ang eta	eIndex: 569 ₀ entries, 0 to columns (total 33 column	⁵⁶⁸ 0.2416 s):		0.1860	0.2
Ħ	Column 0.42450	Non _ō Null Count	Dtype 	0.2430	0.3
10 132 364 5 6676866716111618161716 3 8 4 · 4 0 5 20618273748516	id 0.86630 diagnosis	569 ₀ nop _ē gull 569 non-null	int64 object	0.2575	0.6
2 84	radius_mean.20500 texture_mean	569 ₀ .40ō0ull 569 non-null	float64 float64	0.1625	0.2
4 5	perimeter_mean area_mean	569 non-null 569 non-null	float64 float64	• • •	
54 70	smoothness ₀ mean compactness_mean	569 ₀ .4107 569 non-null	float64 float64	0.2216	0.2
55	concavity_meag ₂₂₀ concave points_mean	569 ₀ .32 <u>1</u> 5ull 569 non-null	float64 float64	0.1628	0.2
10 10 10 10 10 10 10 10 10 10 10 10 10 1	symmetry_mgan fractal_dimension_mean	569 ₀ .34 ₀ 3ull 569 non-null	float64 float64	0.1418	0.2
[2 57 [3 87	radius_se 0.86810 texture_se	569 ₀ 9387 569 non-null	float64 float64	0.2650	0.4
14 58 15 71	perimeter_§e _{.06444} area_se	⁵⁶⁹ 0.00ō0ull 569 non-null	float64 float64	0.0000	0.2
16 17	smoothness_se	569 non-null	float64 float64		
18	compactness se fractal dimension_worst concavity_se 0.11890	569 non-null 569 non-null	float64		
L9	concave points_se.08902	569 non-null	float64 float64		
20 21	symmetry_se 0.08758 fractal_dimension_se_300	569 non-null	float64		
22	radius_worst 0.07678	569 non-null	float64		
23	texture_worst	569 non-null	float64		
<u> </u>	perimeter_worst 0.07115 area_worst 0.0637	569 non-null	float64 float64		
5	smoothness_worst 0.06637	569 non-null	float64		
37 57	compactness_worsh 12400	268 non-Valt	float64		
445566778889	concavity_worst 0.07039	569 non-null	float64		
29 20	concave points_worst	569 non-null 569 non-null	float64 float64		
869 31	<pre>symmetry_worst rows x 33 columns]> fractal_dimension_worst</pre>	569 non-null	float64		
32	Unnamed: 32	0 non-null	float64		
typ	es: float64(31), int64(1) ry usage: 146.8+ KB	, object(1)			

In [8]:

df.isnull().sum()

Out[8]:

id	0
diagnosis	0
radius_mean	0
texture_mean	0
perimeter_mean	0
area_mean	0
smoothness_mean	0
compactness_mean	0
concavity_mean	0
concave points_mean	0
symmetry_mean	0
<pre>fractal_dimension_mean</pre>	0
radius_se	0
texture_se	0
perimeter_se	0
area_se	0
smoothness_se	0
compactness_se	0
concavity_se	0
concave points_se	0
symmetry_se	0
<pre>fractal_dimension_se</pre>	0
radius_worst	0
texture_worst	0
perimeter_worst	0
area_worst	0
smoothness_worst	0
compactness_worst	0
concavity_worst	0
concave points_worst	0
symmetry_worst	0
<pre>fractal_dimension_worst</pre>	0
Unnamed: 32	569

dtype: int64

In [9]:

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

Out[9]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothn
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	M	20.60	29.33	140.10	1265.0	
568	92751	В	7.76	24.54	47.92	181.0	

569 rows × 32 columns

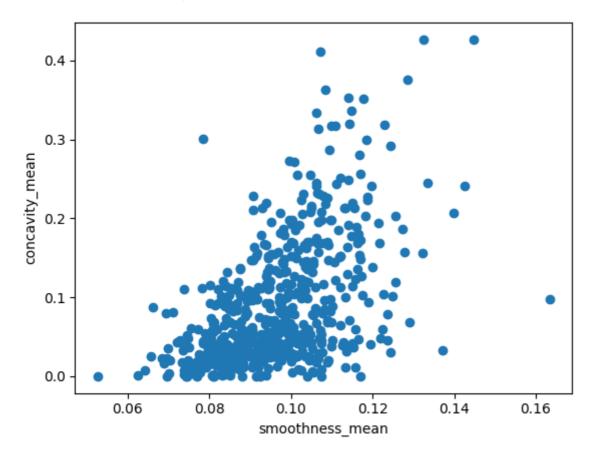
STEP-3:Data Visualization

In [10]:

```
plt.scatter(df["smoothness_mean"],df["concavity_mean"])
plt.xlabel("smoothness_mean")
plt.ylabel("concavity_mean")
```

Out[10]:

Text(0, 0.5, 'concavity_mean')



STEP-4:Data Modelling

In [11]:

```
from sklearn.cluster import KMeans
km=KMeans()
km
```

Out[11]:

KMeans()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

In [12]:

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

C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(

Out[12]:

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

In [13]:

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

Out[13]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothnes
0	842302	М	17.99	10.38	122.80	1001.0	_
1	842517	M	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	

5 rows × 34 columns

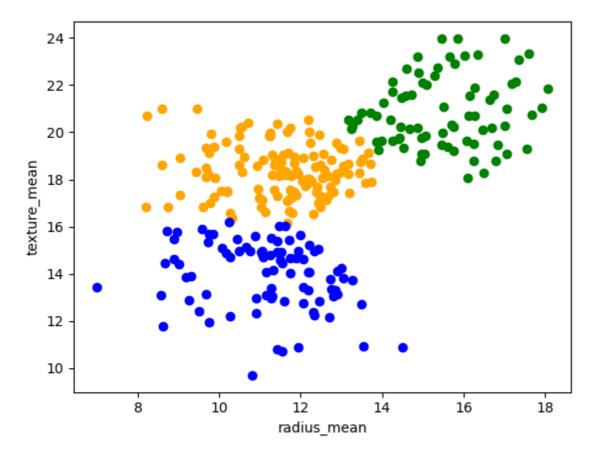
→

In [16]:

```
df1=df[df.cluster==0]
df2=df[df.cluster==1]
df3=df[df.cluster==2]
plt.scatter(df1["radius_mean"],df1["texture_mean"],color="Orange")
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[16]:

Text(0, 0.5, 'texture_mean')



In [17]:

```
km.cluster_centers_
```

Out[17]:

In [20]:

```
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
scaler.fit(df[["concavity_mean"]])
df["concavity_mean"]=scaler.transform(df[["concavity_mean"]])
df.head()
```

Out[20]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothnes
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	

5 rows × 34 columns

→

In [21]:

```
scaler=MinMaxScaler()
scaler.fit(df[["smoothness_mean"]])
df["smoothness_mean"]=scaler.transform(df[["smoothness_mean"]])
df.head()
```

Out[21]:

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

5 rows × 34 columns

→

In [22]:

km=KMeans()

In [23]:

```
y_predicted=km.fit_predict(df[["smoothness_mean","concavity_mean"]])
y_predicted
```

C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(

Out[23]:

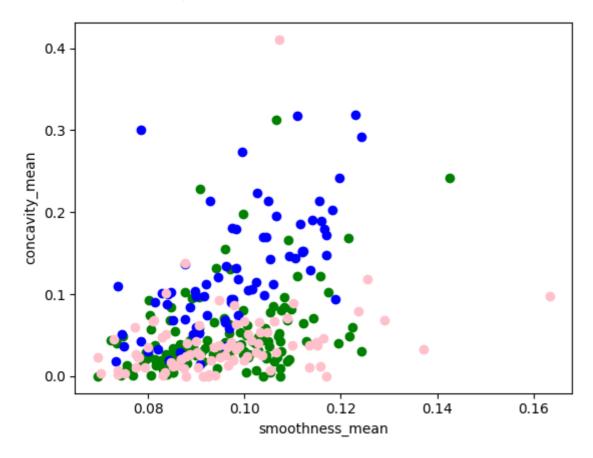
```
array([5, 0, 6, 6, 1, 6, 0, 4, 6, 6, 2, 0, 1, 0, 6, 6, 0, 6, 7, 3, 3, 3,
       1, 0, 7, 6, 7, 7, 7, 0, 1, 7, 6, 7, 7, 7, 7, 3, 3, 7, 2, 4, 1, 7,
       0, 6, 2, 4, 3, 2, 2, 2, 6, 3, 3, 7, 7, 2, 3, 4, 4, 1, 2, 4, 4,
       3, 2, 5, 3, 0, 0, 7, 0, 3, 0, 4, 7, 5, 3, 4, 7, 5, 6, 3, 7,
       0, 4, 3,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

In [19]:

```
df1=df[df.cluster==0]
df2=df[df.cluster==1]
df3=df[df.cluster==2]
plt.scatter(df1["smoothness_mean"],df1["concavity_mean"],color='green')
plt.scatter(df2["smoothness_mean"],df2["concavity_mean"],color='blue')
plt.scatter(df3["smoothness_mean"],df3["concavity_mean"],color='pink')
plt.xlabel("smoothness_mean")
plt.ylabel("concavity_mean")
```

Out[19]:

Text(0, 0.5, 'concavity_mean')



In [24]:

```
km.cluster_centers_
```

Out[24]:

In [29]:

k_rng=range(1,10)
sse=[]

In [30]:

```
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\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
  warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
  warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
  warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
  warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(
C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(
[17561.326763272406, 9403.76824064771, 6745.615775035164, 5152.17519709165
8, 4334.319755681291, 3659.4959704257726, 3169.286852219091, 2820.01019814
```

31917, 2572.518235642421]

C:\Users\thara\AppData\Local\Programs\Python\Python310\lib\site-packages\s
klearn\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` explicit
ly to suppress the warning
 warnings.warn(

Out[30]:

Text(0, 0.5, 'Sum of Squared Error')

