

Problem Statement : Predictive study using the breastcancer diagnostic data set

DATA COLLECTION

```
In [1]: 1 import pandas as pd
        2 from matplotlib import pyplot as plt
        3 %matplotlib inline
```

```
In [2]: 1 df=pd.read_csv(r"C:\Users\Dell\Downloads\BreastCancerPrediction.csv")
        2 df
```

Out[2]:

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

569 rows × 33 columns



In [3]:

```
1 df.head()
```

Out[3]:

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

5 rows × 33 columns



In [4]:

```
1 df.tail()
```

Out[4]:

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

5 rows × 33 columns



In [5]:

```
1 df.shape
```

Out[5]: (569, 33)

In [6]:

```
1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 33 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   id                                     569 non-null    int64
1   diagnosis                             569 non-null    object
2   radius_mean                           569 non-null    float64
3   texture_mean                          569 non-null    float64
4   perimeter_mean                        569 non-null    float64
5   area_mean                             569 non-null    float64
6   smoothness_mean                       569 non-null    float64
7   compactness_mean                      569 non-null    float64
8   concavity_mean                        569 non-null    float64
9   concave points_mean                   569 non-null    float64
10  symmetry_mean                         569 non-null    float64
11  fractal_dimension_mean                 569 non-null    float64
12  radius_se                              569 non-null    float64
13  texture_se                             569 non-null    float64
14  perimeter_se                           569 non-null    float64
15  area_se                                569 non-null    float64
16  smoothness_se                          569 non-null    float64
17  compactness_se                         569 non-null    float64
18  concavity_se                           569 non-null    float64
19  concave points_se                      569 non-null    float64
20  symmetry_se                            569 non-null    float64
21  fractal_dimension_se                   569 non-null    float64
22  radius_worst                           569 non-null    float64
23  texture_worst                          569 non-null    float64
24  perimeter_worst                        569 non-null    float64
25  area_worst                             569 non-null    float64
26  smoothness_worst                       569 non-null    float64
27  compactness_worst                      569 non-null    float64
28  concavity_worst                        569 non-null    float64
29  concave points_worst                   569 non-null    float64
30  symmetry_worst                         569 non-null    float64
31  fractal_dimension_worst                 569 non-null    float64
32  Unnamed: 32                             0 non-null      float64
dtypes: float64(31), int64(1), object(1)
memory usage: 146.8+ KB
```

In [7]:

```
1 df.describe
```

```

Out[7]: <bound method NDFrame.describe of
ure_mean  perimeter_mean  area_mean  \
0      842302      M      17.99      10.38      122.80      1001.
0
1      842517      M      20.57      17.77      132.90      1326.
0
2      84300903      M      19.69      21.25      130.00      1203.
0
3      84348301      M      11.42      20.38      77.58      386.
1
4      84358402      M      20.29      14.34      135.10      1297.
0
..      ...      ...      ...      ...      ...
...
564      926424      M      21.56      22.39      142.00      1479.
0
565      926682      M      20.13      28.25      131.20      1261.
0
566      926954      M      16.60      28.08      108.30      858.
1
567      927241      M      20.60      29.33      140.10      1265.
0
568      92751      B      7.76      24.54      47.92      181.
0

      smoothness_mean  compactness_mean  concavity_mean  concave points_mean
\
0      0.11840      0.27760      0.30010      0.14710
1      0.08474      0.07864      0.08690      0.07017
2      0.10960      0.15990      0.19740      0.12790
3      0.14250      0.28390      0.24140      0.10520
4      0.10030      0.13280      0.19800      0.10430
..      ...      ...      ...      ...
564      0.11100      0.11590      0.24390      0.13890
565      0.09780      0.10340      0.14400      0.09791
566      0.08455      0.10230      0.09251      0.05302
567      0.11780      0.27700      0.35140      0.15200
568      0.05263      0.04362      0.00000      0.00000

      ... texture_worst  perimeter_worst  area_worst  smoothness_worst  \
0      ...      17.33      184.60      2019.0      0.16220
1      ...      23.41      158.80      1956.0      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
568      ...      30.37      59.16      268.6      0.08996

      compactness_worst  concavity_worst  concave points_worst  symmetry_wors
t \
0      0.66560      0.7119      0.2654      0.460
1
1      0.18660      0.2416      0.1860      0.275

```

0				
2	0.42450	0.4504	0.2430	0.361
3				
3	0.86630	0.6869	0.2575	0.663
8				
4	0.20500	0.4000	0.1625	0.236
4				
..	
...				
564	0.21130	0.4107	0.2216	0.206
0				
565	0.19220	0.3215	0.1628	0.257
2				
566	0.30940	0.3403	0.1418	0.221
8				
567	0.86810	0.9387	0.2650	0.408
7				
568	0.06444	0.0000	0.0000	0.287
1				

	fractal_dimension_worst	Unnamed: 32
0	0.11890	NaN
1	0.08902	NaN
2	0.08758	NaN
3	0.17300	NaN
4	0.07678	NaN
..
564	0.07115	NaN
565	0.06637	NaN
566	0.07820	NaN
567	0.12400	NaN
568	0.07039	NaN

[569 rows x 33 columns]>

```
In [8]: 1 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
fractal_dimension_mean                   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
fractal_dimension_se                     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
fractal_dimension_worst                   0
Unnamed: 32                              569
dtype: int64
```

In [9]: 1 df.drop(['Unnamed: 32'],axis=1)

Out[9]:

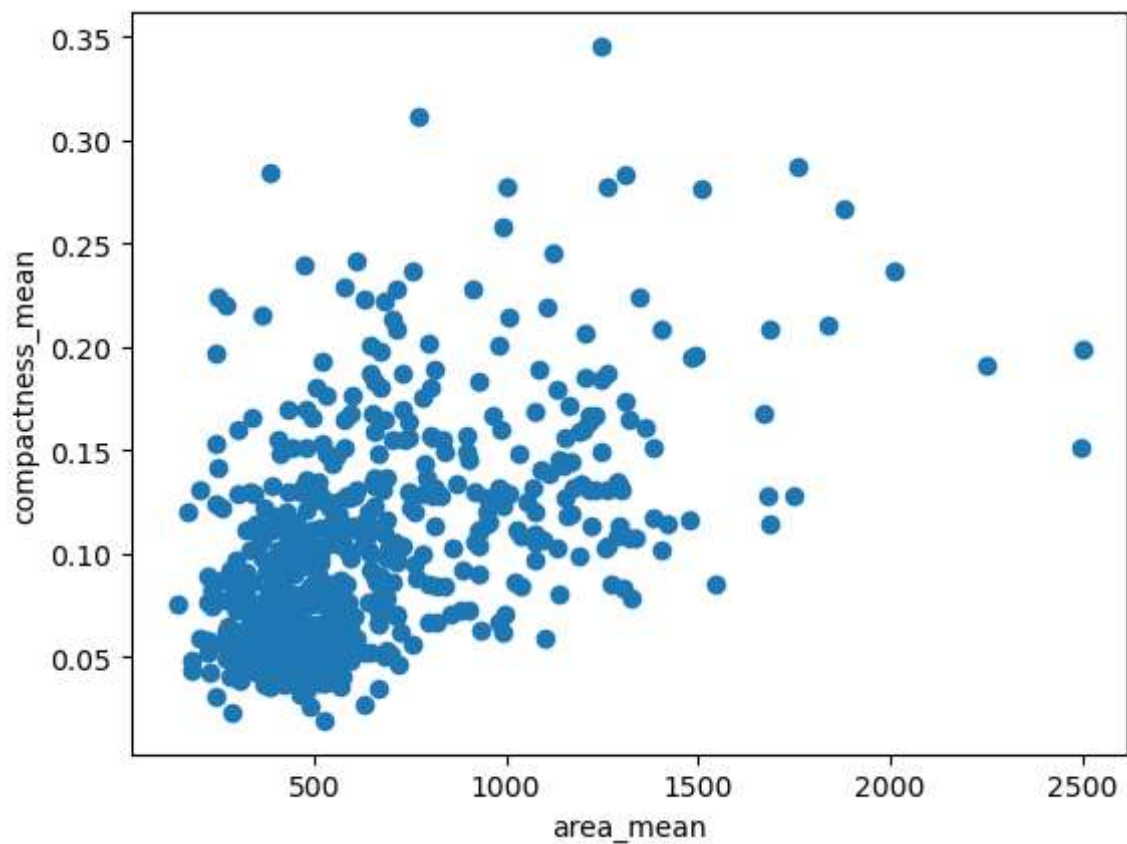
	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness
0	842302	M	17.99	10.38	122.80	1001.0	(
1	842517	M	20.57	17.77	132.90	1326.0	C
2	84300903	M	19.69	21.25	130.00	1203.0	C
3	84348301	M	11.42	20.38	77.58	386.1	C
4	84358402	M	20.29	14.34	135.10	1297.0	C
...	
564	926424	M	21.56	22.39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

569 rows × 32 columns




```
In [13]: 1 plt.scatter(df["area_mean"],df["compactness_mean"])
2 plt.xlabel("area_mean")
3 plt.ylabel("compactness_mean")
```

Out[13]: Text(0, 0.5, 'compactness_mean')



```
In [14]: 1 from sklearn.cluster import KMeans
```

```
In [15]: 1 km=KMeans()
2 km
```

Out[15]:

▼ KMeans

KMeans()

```
In [16]: 1 y_predicted=km.fit_predict(df[["area_mean","compactness_mean"]])
        2 y_predicted
```

C:\ProgramData\anaconda3\lib\site-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(

C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:1382: 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=3.

warnings.warn(

```
Out[16]: array([6, 0, 0, 1, 0, 1, 6, 7, 7, 1, 3, 3, 0, 3, 7, 3, 3, 3, 0, 7, 7, 4,
                3, 0, 6, 6, 3, 6, 3, 6, 6, 1, 6, 0, 3, 6, 7, 7, 3, 7, 7, 1, 6, 7,
                7, 6, 4, 7, 1, 7, 1, 7, 1, 6, 3, 1, 0, 3, 7, 4, 4, 4, 3, 4, 7, 3,
                4, 1, 4, 7, 0, 4, 6, 7, 1, 3, 7, 6, 0, 7, 1, 7, 2, 0, 1, 6, 3, 6,
                1, 3, 3, 3, 7, 7, 3, 0, 1, 4, 1, 3, 7, 4, 1, 4, 4, 7, 1, 1, 2, 1,
                4, 1, 7, 4, 4, 1, 4, 3, 3, 6, 1, 6, 2, 3, 7, 7, 7, 0, 3, 0, 1, 3,
                3, 3, 6, 7, 1, 1, 3, 1, 4, 3, 1, 7, 1, 1, 1, 3, 3, 7, 7, 4, 4, 1,
                7, 1, 6, 6, 1, 1, 1, 0, 0, 1, 2, 3, 1, 6, 6, 3, 1, 7, 3, 1, 4, 4,
                4, 3, 7, 7, 5, 0, 3, 1, 3, 4, 6, 1, 1, 1, 7, 7, 4, 1, 3, 7, 7, 6,
                0, 3, 1, 6, 2, 7, 1, 3, 4, 6, 7, 3, 0, 1, 5, 6, 7, 7, 1, 4, 0, 0,
                7, 7, 4, 3, 7, 3, 4, 3, 7, 7, 6, 1, 1, 0, 4, 7, 2, 0, 7, 6, 7, 1,
                1, 7, 0, 4, 7, 7, 4, 1, 0, 1, 0, 6, 0, 7, 0, 3, 3, 3, 0, 6, 6, 3,
                6, 0, 4, 7, 7, 4, 7, 1, 2, 4, 6, 1, 1, 6, 7, 7, 0, 1, 0, 3, 7, 7,
                1, 7, 1, 1, 3, 3, 7, 1, 7, 7, 1, 1, 7, 4, 0, 1, 0, 4, 1, 1, 7, 4,
                7, 7, 1, 3, 7, 1, 4, 1, 1, 6, 4, 1, 4, 0, 7, 0, 1, 7, 7, 1, 3, 3,
                3, 7, 1, 1, 1, 6, 7, 6, 4, 2, 3, 4, 1, 0, 1, 4, 1, 3, 1, 1, 1, 3,
                2, 3, 1, 1, 7, 7, 4, 4, 7, 7, 7, 3, 7, 0, 0, 1, 2, 2, 3, 3, 0, 0,
                7, 3, 4, 7, 7, 1, 1, 1, 1, 1, 7, 3, 1, 7, 1, 0, 4, 4, 3, 0, 1, 7,
                7, 7, 1, 1, 6, 1, 7, 7, 1, 1, 3, 7, 6, 1, 1, 1, 4, 3, 3, 1, 4, 3,
                7, 1, 1, 3, 1, 7, 4, 4, 4, 1, 1, 7, 3, 1, 0, 6, 3, 7, 7, 7, 7, 7,
                1, 6, 7, 4, 6, 1, 6, 3, 3, 0, 1, 0, 1, 3, 7, 7, 1, 7, 7, 4, 6, 5,
                3, 1, 7, 7, 7, 4, 6, 1, 4, 1, 3, 1, 1, 7, 7, 7, 1, 3, 1, 7, 7, 7,
                3, 1, 3, 0, 1, 6, 1, 6, 6, 1, 7, 3, 1, 1, 6, 0, 3, 7, 1, 2, 4, 4,
                1, 1, 3, 3, 1, 3, 7, 3, 3, 1, 6, 0, 7, 7, 4, 2, 1, 7, 4, 4, 7, 1,
                7, 1, 1, 1, 7, 0, 1, 0, 7, 1, 4, 4, 1, 3, 3, 7, 7, 7, 4, 4, 4, 1,
                1, 1, 7, 4, 7, 4, 4, 4, 3, 1, 7, 1, 3, 0, 2, 0, 6, 0, 4])
```

```
In [17]: 1 df["cluster"]=y_predicted
```

```
In [19]: 1 df.head()
```

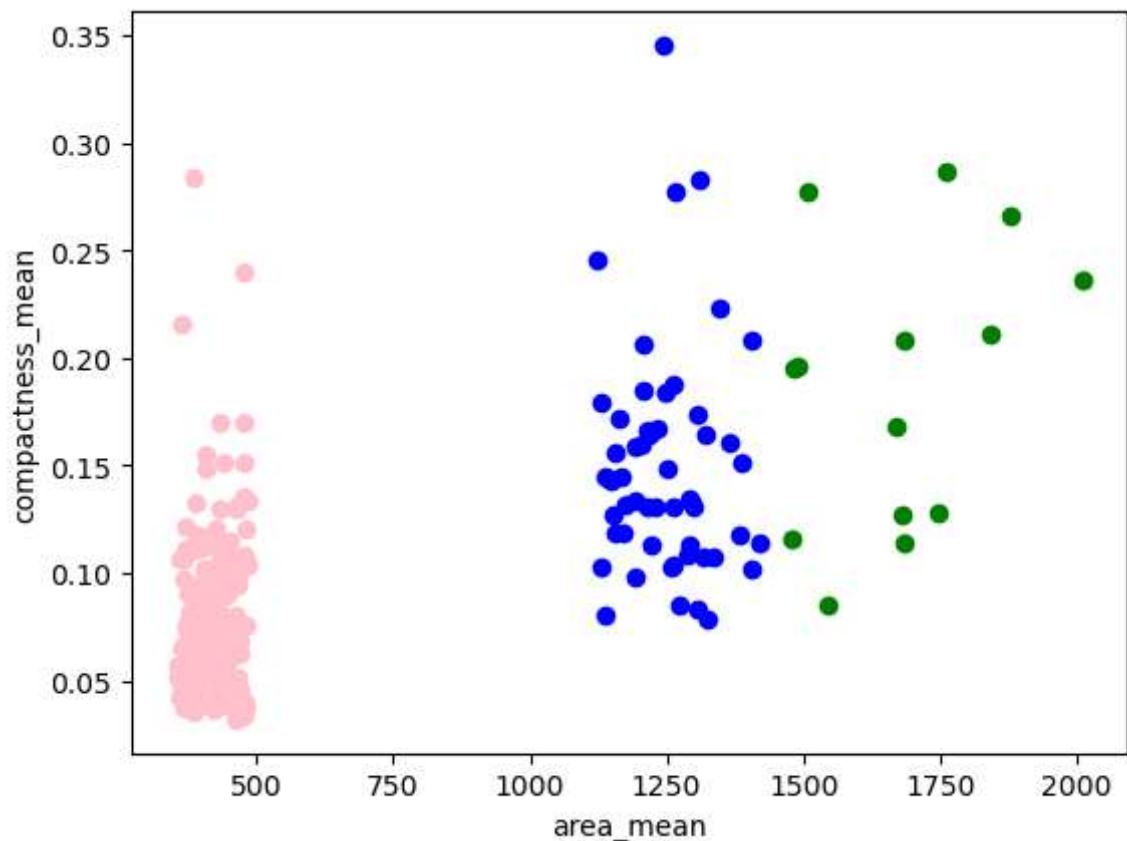
```
Out[19]:
```

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

5 rows × 34 columns

```
In [20]: 1 df1=df[df.cluster==0]
2 df2=df[df.cluster==1]
3 df3=df[df.cluster==2]
4 plt.scatter(df1["area_mean"],df1["compactness_mean"],color='blue')
5 plt.scatter(df2["area_mean"],df2["compactness_mean"],color='pink')
6 plt.scatter(df3["area_mean"],df3["compactness_mean"],color='green')
7 plt.xlabel("area_mean")
8 plt.ylabel("compactness_mean")
```

```
Out[20]: Text(0, 0.5, 'compactness_mean')
```



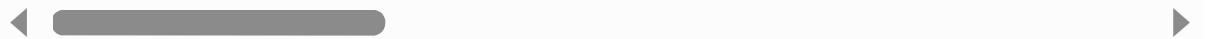
```
In [21]: 1 from sklearn.preprocessing import MinMaxScaler
2 scaler=MinMaxScaler()
3 scaler.fit(df[["compactness_mean"]])
4 df["compactness_mean"]=scaler.transform(df[["compactness_mean"]])
```

```
In [22]: 1 df.head()
```

Out[22]:

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

5 rows × 34 columns



```
In [23]: 1 scaler=MinMaxScaler()
2 scaler.fit(df[["area_mean"]])
3 df["area_mean"]=scaler.transform(df[["area_mean"]])
```

```
In [24]: 1 df.head()
```

Out[24]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_n
0	842302	M	17.99	10.38	122.80	0.363733	0.1
1	842517	M	20.57	17.77	132.90	0.501591	0.0
2	84300903	M	19.69	21.25	130.00	0.449417	0.1
3	84348301	M	11.42	20.38	77.58	0.102906	0.1
4	84358402	M	20.29	14.34	135.10	0.489290	0.1

5 rows × 34 columns



```
In [25]: 1 km=KMeans()
```

```
In [26]: 1 y_predicted=km.fit_predict(df[["area_mean","compactness_mean"]])
        2 y_predicted
```

```
C:\ProgramData\anaconda3\lib\site-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(
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1382: 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=3.
```

```
warnings.warn(
```

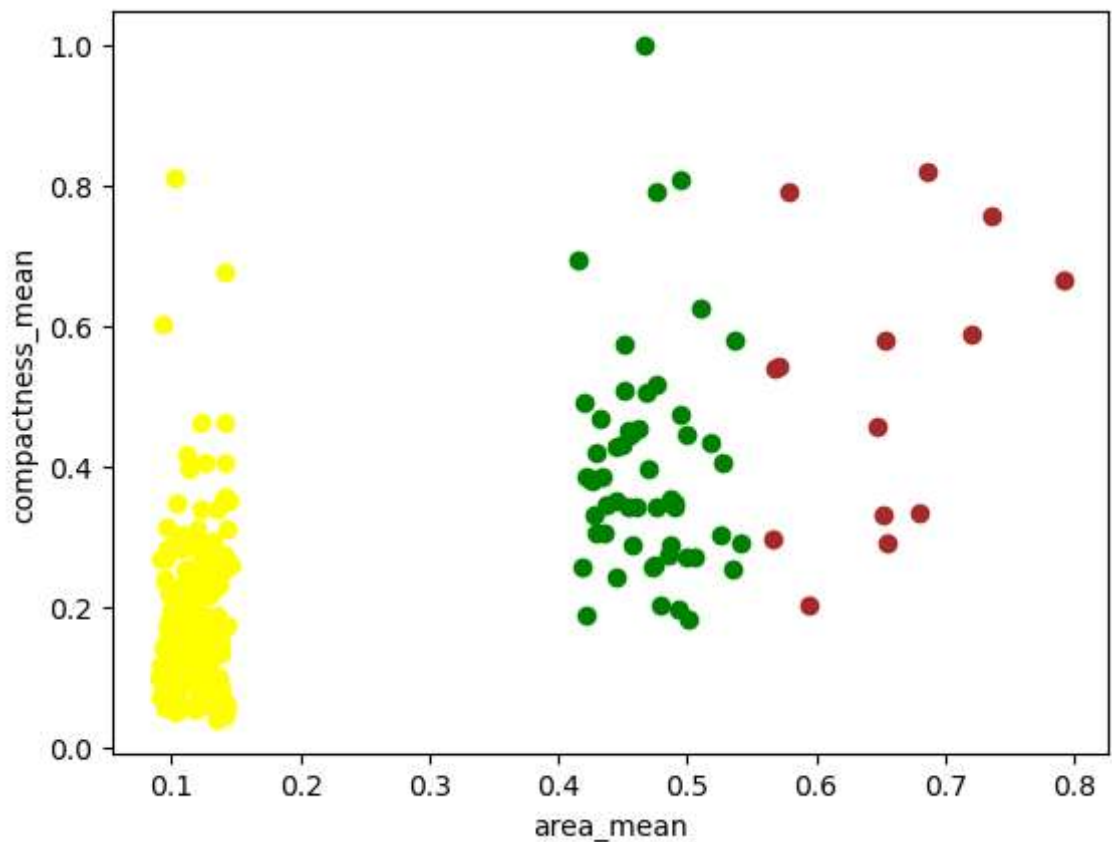
```
Out[26]: array([2, 1, 3, 2, 1, 5, 1, 5, 5, 2, 6, 5, 2, 6, 2, 5, 6, 2, 1, 6, 4, 0,
                2, 1, 5, 2, 5, 1, 5, 1, 3, 5, 5, 3, 5, 5, 4, 0, 6, 4, 6, 4, 2, 5,
                4, 3, 0, 4, 4, 6, 0, 6, 0, 1, 6, 0, 1, 5, 0, 0, 0, 0, 5, 0, 4, 5,
                0, 0, 4, 0, 1, 4, 5, 4, 0, 6, 4, 2, 2, 4, 4, 5, 7, 3, 0, 1, 4, 1,
                4, 5, 6, 6, 0, 6, 5, 1, 0, 0, 0, 4, 6, 0, 0, 4, 0, 5, 4, 0, 7, 0,
                0, 4, 2, 4, 0, 0, 4, 5, 5, 6, 0, 1, 7, 4, 6, 6, 6, 1, 5, 3, 4, 5,
                5, 6, 1, 0, 0, 0, 5, 4, 0, 1, 4, 4, 0, 4, 5, 4, 4, 6, 6, 4, 5, 0,
                6, 0, 5, 6, 0, 0, 4, 1, 3, 4, 1, 6, 0, 6, 5, 6, 0, 6, 5, 0, 0, 0,
                4, 5, 0, 0, 7, 2, 6, 0, 4, 0, 1, 0, 0, 0, 2, 0, 0, 4, 5, 0, 4, 1,
                1, 4, 0, 1, 7, 5, 4, 6, 0, 6, 5, 6, 1, 0, 7, 1, 4, 5, 4, 0, 1, 1,
                6, 4, 0, 5, 0, 6, 0, 4, 6, 5, 5, 0, 0, 1, 0, 6, 3, 1, 4, 1, 6, 0,
                4, 6, 3, 0, 0, 4, 0, 0, 3, 0, 3, 1, 1, 4, 3, 2, 2, 5, 1, 6, 1, 6,
                1, 1, 4, 6, 0, 4, 0, 0, 3, 0, 1, 0, 0, 1, 0, 6, 1, 0, 1, 5, 4, 0,
                4, 0, 4, 0, 5, 6, 6, 0, 6, 6, 0, 0, 6, 0, 3, 4, 3, 0, 0, 0, 0, 0,
                0, 0, 0, 6, 6, 0, 0, 0, 0, 1, 5, 0, 4, 1, 4, 3, 0, 6, 6, 0, 5, 5,
                5, 4, 0, 0, 0, 1, 4, 1, 0, 3, 4, 4, 4, 1, 0, 4, 0, 6, 0, 4, 0, 2,
                7, 6, 0, 4, 4, 6, 0, 0, 0, 6, 6, 6, 6, 1, 3, 0, 1, 3, 5, 6, 3, 1,
                6, 5, 5, 0, 6, 2, 4, 0, 4, 4, 6, 6, 0, 6, 4, 1, 0, 0, 5, 3, 4, 6,
                4, 4, 0, 0, 2, 0, 6, 4, 0, 0, 6, 6, 1, 0, 0, 0, 0, 6, 6, 0, 0, 5,
                0, 0, 0, 5, 4, 4, 0, 0, 4, 0, 0, 0, 2, 4, 3, 1, 6, 4, 6, 6, 6, 6,
                4, 1, 6, 0, 1, 4, 1, 6, 6, 1, 4, 1, 0, 6, 0, 6, 4, 0, 0, 0, 1, 7,
                6, 0, 6, 4, 4, 0, 2, 5, 0, 0, 6, 0, 4, 6, 4, 6, 0, 5, 0, 6, 4, 6,
                5, 5, 6, 1, 4, 6, 0, 6, 1, 0, 0, 6, 4, 0, 1, 3, 5, 5, 4, 3, 2, 2,
                4, 4, 6, 5, 4, 6, 5, 6, 6, 0, 1, 1, 5, 4, 4, 7, 0, 4, 0, 0, 4, 0,
                4, 4, 4, 4, 6, 1, 4, 3, 4, 5, 0, 4, 4, 4, 6, 6, 4, 6, 0, 0, 0, 0,
                0, 0, 0, 0, 0, 0, 0, 0, 5, 4, 4, 0, 2, 3, 1, 1, 1, 2, 0])
```

```

In [27]: 1 df1=df[df.cluster==0]
          2 df2=df[df.cluster==1]
          3 df3=df[df.cluster==2]
          4 plt.scatter(df1["area_mean"],df1["compactness_mean"],color='green')
          5 plt.scatter(df2["area_mean"],df2["compactness_mean"],color='yellow')
          6 plt.scatter(df3["area_mean"],df3["compactness_mean"],color='brown')
          7 plt.xlabel("area_mean")
          8 plt.ylabel("compactness_mean")

```

Out[27]: Text(0, 0.5, 'compactness_mean')



```

In [28]: 1 km.cluster_centers_

```

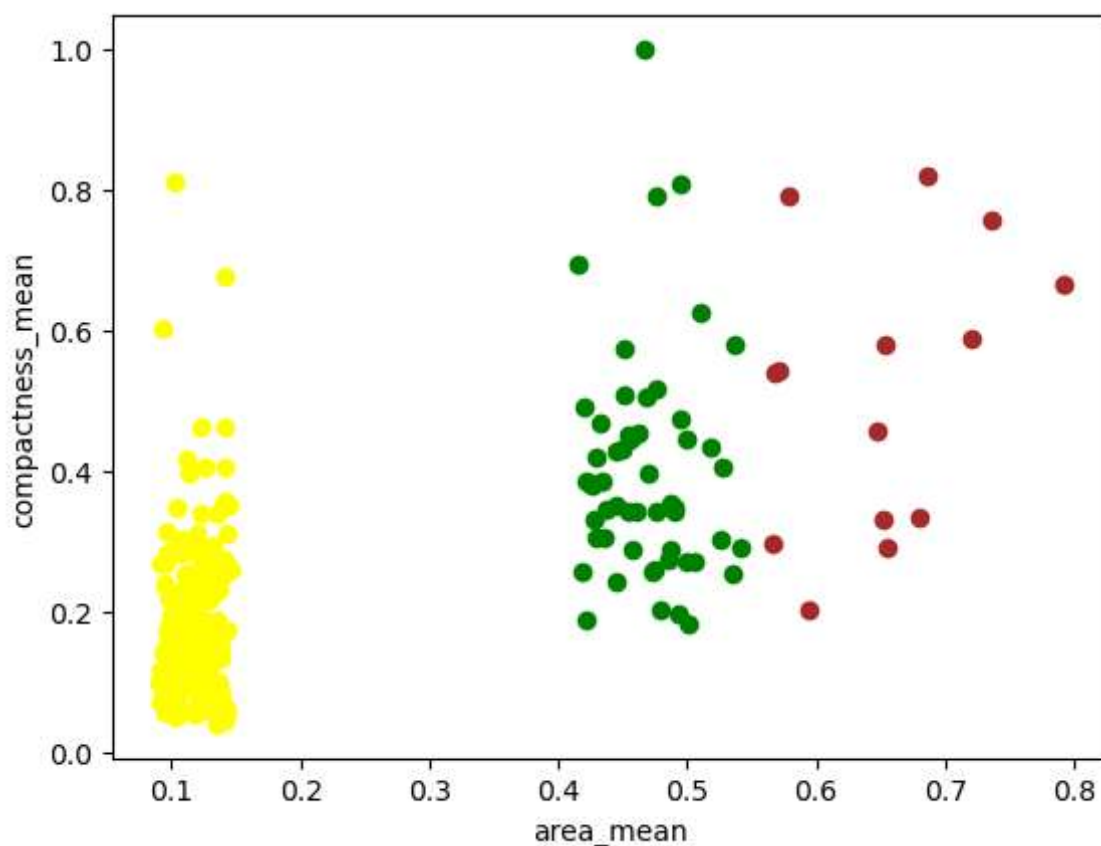
Out[28]: array([[0.10629362, 0.1188301],
 [0.4229776 , 0.29085592],
 [0.27262948, 0.68264524],
 [0.49465699, 0.4702779],
 [0.13942697, 0.28139861],
 [0.22385074, 0.42539472],
 [0.21320951, 0.16216208],
 [0.7843761 , 0.63141321]])

```

In [29]: 1 df1=df[df.cluster==0]
          2 df2=df[df.cluster==1]
          3 df3=df[df.cluster==2]
          4 plt.scatter(df1["area_mean"],df1["compactness_mean"],color='green')
          5 plt.scatter(df2["area_mean"],df2["compactness_mean"],color='yellow')
          6 plt.scatter(df3["area_mean"],df3["compactness_mean"],color='brown')
          7 plt.xlabel("area_mean")
          8 plt.ylabel("compactness_mean")

```

Out[29]: Text(0, 0.5, 'compactness_mean')



```

In [30]: 1 k_rng=range(1,10)
          2 sse=[]

```

In [31]:

```
1 for k in k_rng:
2     km=KMeans(n_clusters=k)
3     km.fit(df[["area_mean", "compactness_mean"]])
4     sse.append(km.inertia_)
5 sse
```



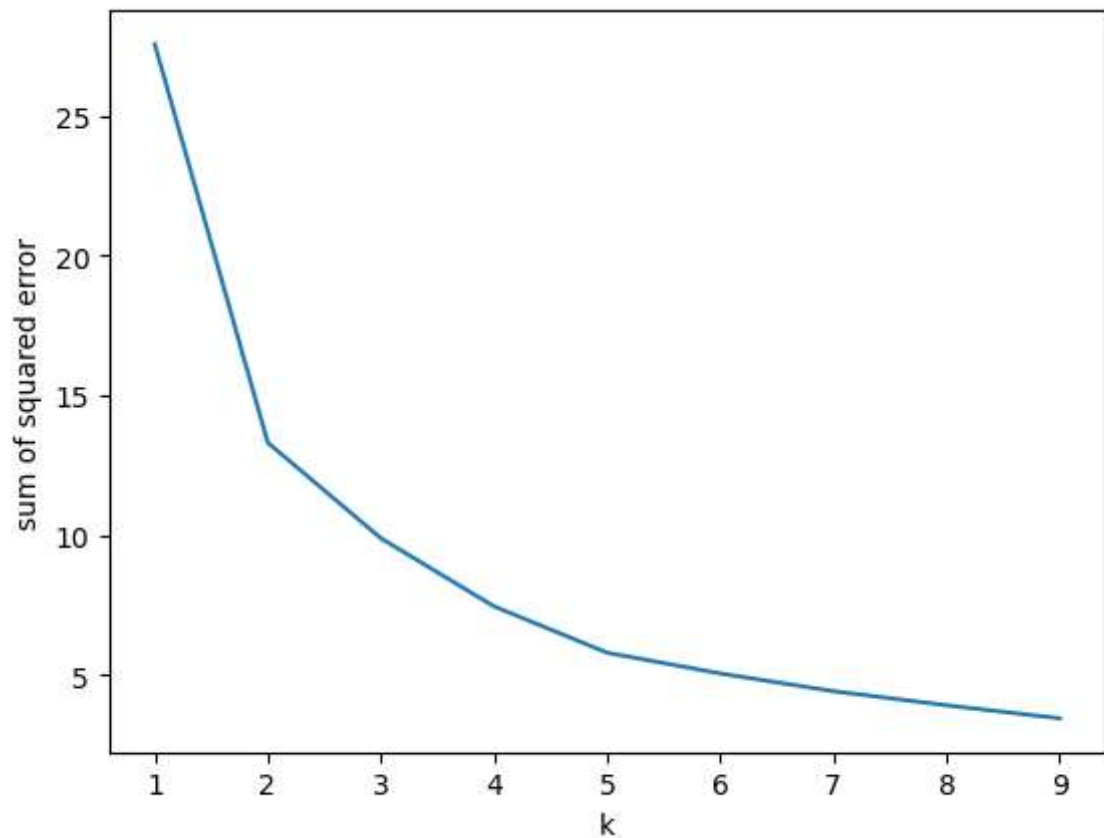
```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
  warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1382: 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=3.
  warnings.warn(
C:\ProgramData\anaconda3\lib\site-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
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C:\ProgramData\anaconda3\lib\site-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
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C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1382: 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=3.
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C:\ProgramData\anaconda3\lib\site-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
```

```
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C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1382:
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warnings.warn(
C:\ProgramData\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:870: F
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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 th
e environment variable OMP_NUM_THREADS=3.
warnings.warn(
```

```
Out[31]: [27.561828426895854,
13.305226534907467,
9.883778331580826,
7.448282131137234,
5.790893797233021,
5.049924168243749,
4.41531873053939,
3.9121245445690334,
3.4447376650665933]
```

```
In [32]: 1 plt.plot(k_rng,sse)
          2 plt.xlabel("k")
          3 plt.ylabel("sum of squared error")
```

```
Out[32]: Text(0, 0.5, 'sum of squared error')
```



CONCLUSION : For the given dataset "BreastCancer Prediction" we used K-means clustering and divide the data in to different clusters for prediction