

PROBLEM STATEMENT:

To perform an analytics report on 100 years of Rainfall data"

importing required libraries

```
In [1]: ▶ import numpy  
import matplotlib.pyplot as plt  
import pygad  
import pandas as pd
```

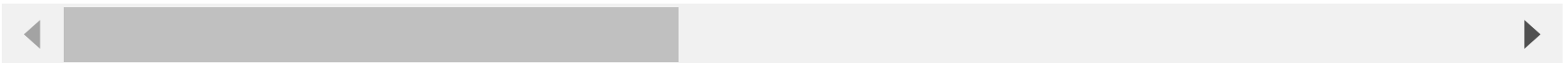
Data collection

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

Out[2]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean
0	842302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.30010
1	842517	M	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.08690
2	84300903	M	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.19740
3	84348301	M	11.42	20.38	77.58	386.1	0.14250	0.28390	0.24140
4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.19800
...
564	926424	M	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390
565	926682	M	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400
566	926954	M	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09250
567	927241	M	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140
568	92751	B	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000

569 rows × 33 columns



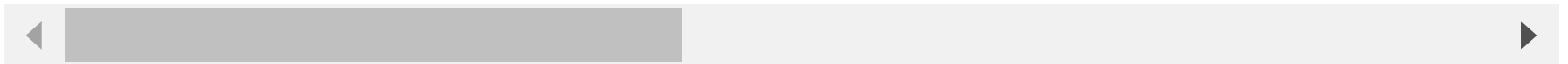
data cleaning

In [3]: `df.head()`

Out[3]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean
0	842302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001
1	842517	M	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869
2	84300903	M	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974
3	84348301	M	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414
4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980

5 rows × 33 columns



In [4]: `df.shape`

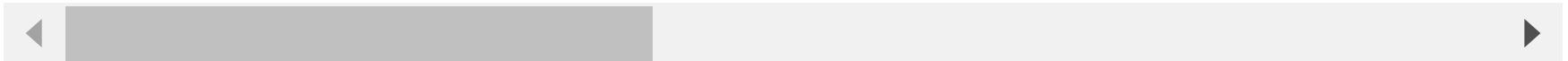
Out[4]: (569, 33)

```
In [5]: df.describe()
```

Out[5]:

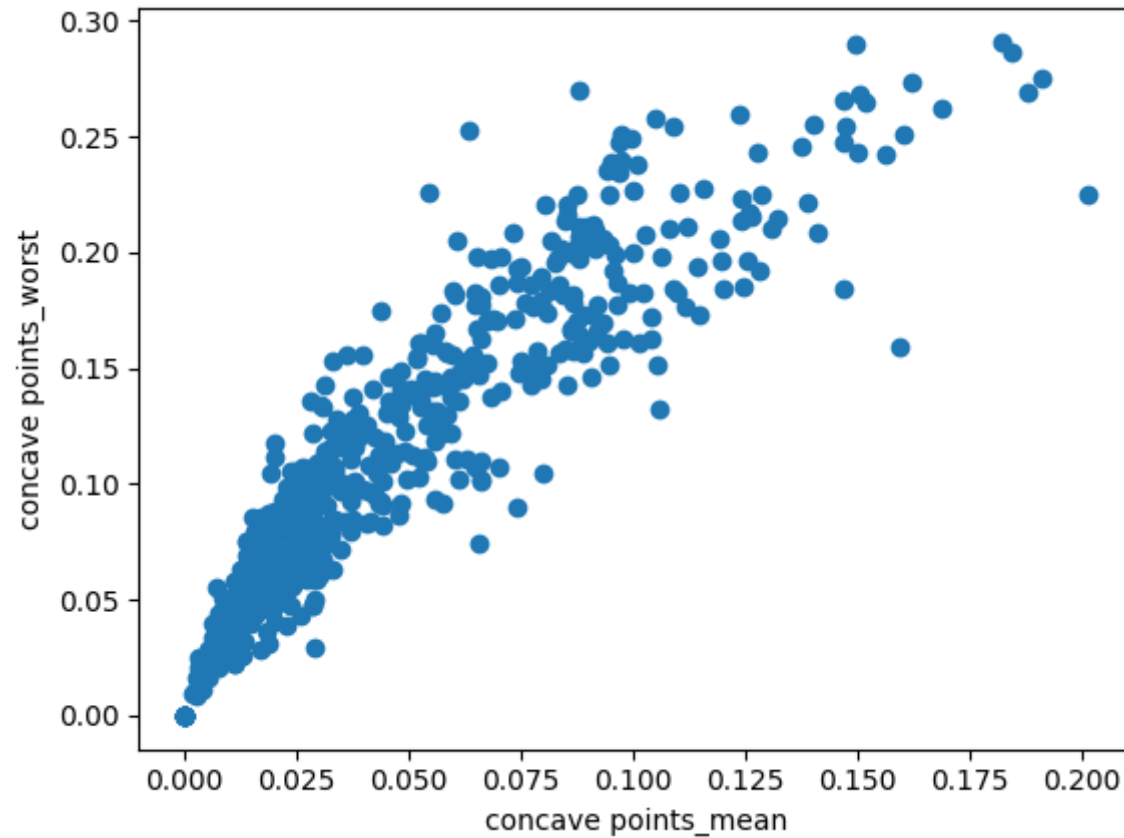
	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean	p
count	5.690000e+02	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	
mean	3.037183e+07	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799	
std	1.250206e+08	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720	
min	8.670000e+03	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000	
25%	8.692180e+05	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560	
50%	9.060240e+05	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540	
75%	8.813129e+06	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700	
max	9.113205e+08	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800	

8 rows × 32 columns



```
In [8]: ▶ plt.scatter(df["concave points_mean"],df["concave points_worst"])  
plt.xlabel("concave points_mean")  
plt.ylabel("concave points_worst")
```

Out[8]: Text(0, 0.5, 'concave points_worst')



```
In [9]: ► from sklearn.cluster import KMeans  
km=KMeans()  
km
```

Out[9]: 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 [10]: ► y_predicted=km.fit_predict(df[["concave points_mean","concave points_worst"]])
y_predicted
```

C:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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(

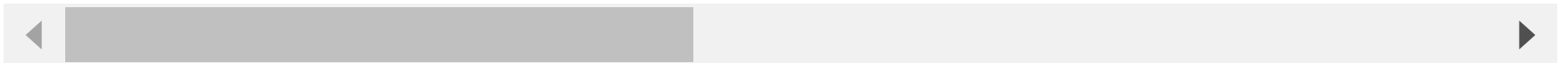
```
Out[10]: array([4, 2, 4, 0, 2, 2, 0, 2, 0, 0, 1, 2, 7, 6, 0, 2, 2, 0, 0, 6, 1, 5,
0, 0, 0, 4, 0, 2, 0, 2, 7, 2, 7, 2, 2, 2, 6, 5, 3, 0, 1, 6, 0, 2,
2, 0, 3, 0, 5, 6, 3, 1, 5, 2, 6, 5, 0, 2, 3, 5, 3, 3, 2, 5, 2, 2,
5, 5, 2, 5, 2, 5, 2, 6, 1, 2, 1, 7, 4, 1, 5, 2, 4, 7, 1, 2, 6, 0,
6, 2, 5, 2, 1, 1, 0, 2, 5, 3, 1, 2, 6, 3, 5, 1, 3, 0, 6, 1, 4, 1,
5, 6, 2, 5, 1, 5, 5, 0, 0, 6, 1, 2, 4, 6, 1, 5, 6, 6, 2, 7, 1, 2,
6, 6, 6, 1, 5, 1, 2, 1, 3, 6, 5, 1, 3, 5, 2, 1, 2, 5, 5, 1, 2, 5,
1, 1, 2, 1, 5, 3, 1, 2, 7, 1, 0, 5, 5, 2, 2, 1, 1, 6, 7, 5, 3, 3,
1, 0, 3, 5, 4, 4, 2, 5, 6, 3, 2, 6, 5, 5, 2, 5, 3, 6, 2, 1, 2, 1,
2, 2, 6, 0, 4, 0, 1, 6, 5, 6, 6, 1, 0, 5, 7, 6, 2, 2, 6, 3, 2, 2,
1, 1, 5, 2, 1, 6, 5, 6, 6, 0, 0, 3, 3, 2, 5, 1, 4, 6, 1, 0, 1, 5,
6, 5, 2, 5, 5, 6, 5, 1, 7, 5, 0, 2, 0, 6, 7, 7, 4, 0, 2, 1, 2, 1,
0, 2, 1, 5, 5, 1, 3, 1, 7, 5, 6, 6, 3, 6, 5, 1, 7, 1, 0, 2, 1, 3,
6, 5, 1, 5, 6, 6, 1, 1, 5, 5, 3, 1, 5, 3, 7, 1, 7, 5, 5, 5, 3, 3,
3, 3, 5, 5, 1, 5, 3, 3, 3, 2, 6, 3, 1, 2, 6, 4, 5, 5, 5, 3, 2, 6,
0, 1, 3, 3, 3, 7, 5, 0, 5, 7, 6, 1, 1, 0, 1, 5, 5, 6, 5, 5, 5, 7,
4, 2, 5, 1, 6, 5, 5, 5, 3, 5, 1, 1, 5, 2, 7, 1, 2, 4, 0, 1, 7, 0,
5, 6, 2, 5, 1, 0, 6, 5, 1, 1, 1, 6, 1, 5, 1, 7, 5, 3, 0, 4, 5, 1,
6, 1, 5, 5, 7, 5, 5, 1, 5, 1, 6, 5, 0, 1, 1, 1, 3, 6, 5, 1, 3, 0,
1, 5, 5, 6, 6, 6, 5, 3, 1, 5, 5, 3, 0, 1, 7, 2, 1, 2, 5, 1, 5, 1,
6, 2, 3, 3, 2, 6, 0, 1, 1, 0, 1, 2, 5, 6, 1, 1, 5, 5, 5, 5, 2, 4,
5, 1, 1, 6, 6, 3, 0, 6, 5, 5, 6, 3, 1, 1, 6, 5, 5, 2, 5, 1, 6, 1,
2, 6, 1, 7, 1, 1, 5, 1, 2, 3, 5, 6, 6, 1, 2, 7, 6, 2, 1, 0, 6, 6,
1, 1, 6, 0, 1, 1, 0, 1, 6, 1, 2, 2, 6, 1, 5, 4, 3, 6, 5, 1, 6, 1,
6, 1, 1, 1, 1, 2, 1, 7, 6, 6, 3, 5, 5, 6, 1, 1, 5, 5, 3, 1, 3, 3,
3, 5, 5, 3, 5, 1, 3, 3, 6, 1, 6, 3, 0, 4, 7, 2, 6, 4, 3])
```

```
In [11]: df["cluster"]=y_predicted  
df.head()
```

Out[11]:

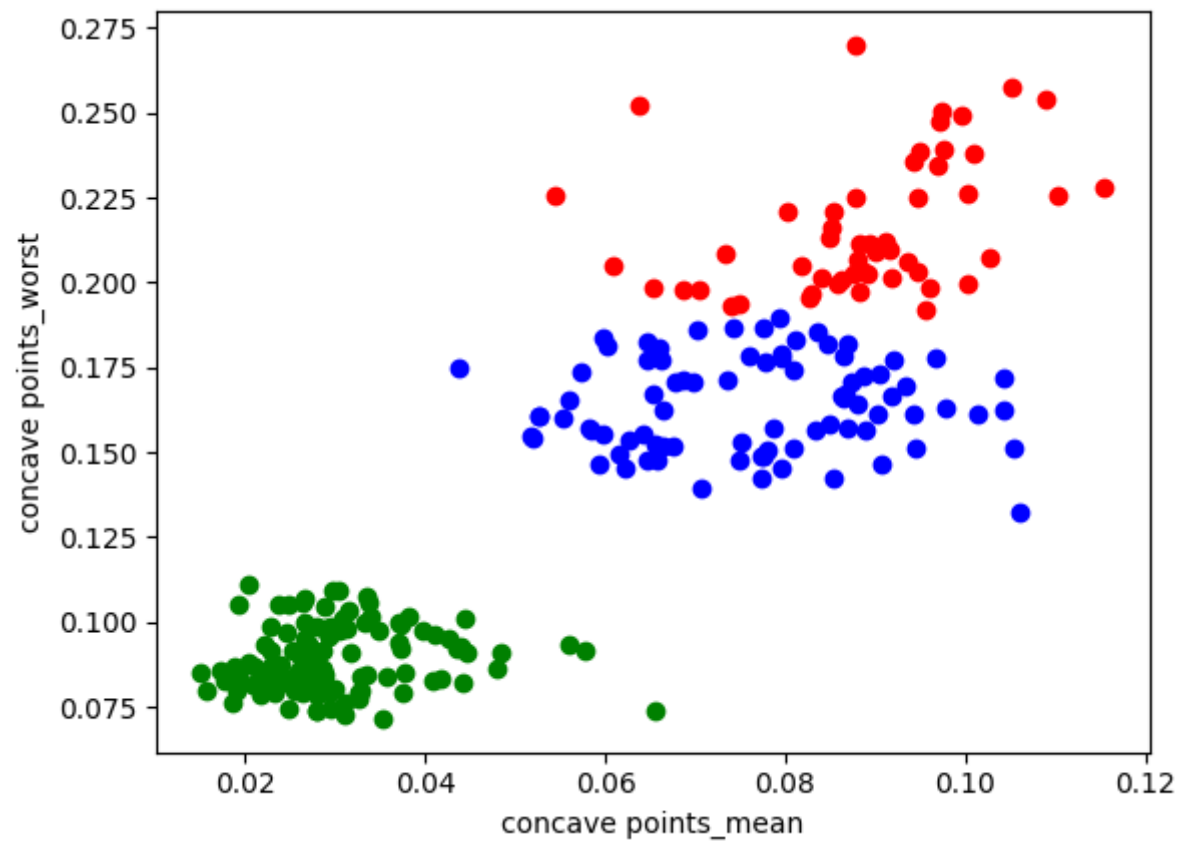
	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean
0	842302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001
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3	84348301	M	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414
4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980

5 rows × 34 columns




```
In [12]: ▶ df1=df[df.cluster==0]
df2=df[df.cluster==1]
df3=df[df.cluster==2]
plt.scatter(df1["concave points_mean"],df1["concave points_worst"],color="red")
plt.scatter(df2["concave points_mean"],df2["concave points_worst"],color="green")
plt.scatter(df3["concave points_mean"],df3["concave points_worst"],color="blue")
plt.xlabel("concave points_mean")
plt.ylabel("concave points_worst")
```

Out[12]: Text(0, 0.5, 'concave points_worst')

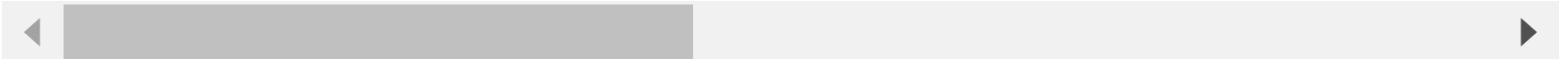


```
In [13]: ▶ from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
scaler.fit(df[["concave points_worst"]])
df["concave points_worst"]=scaler.transform(df[["concave points_worst"]])
df.head()
```

Out[13]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean
0	842302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001
1	842517	M	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869
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3	84348301	M	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414
4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980

5 rows × 10 columns

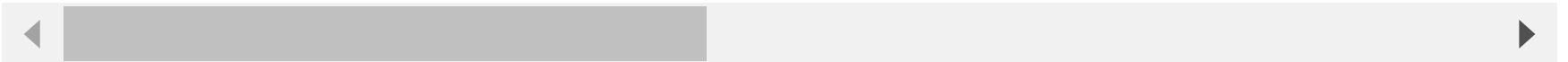


```
In [14]: scaler.fit(df[["concave points_mean"]])  
df["Age"]=scaler.transform(df[["concave points_mean"]])  
df.head()
```

Out[14]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean
0	842302	M	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001
1	842517	M	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869
2	84300903	M	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974
3	84348301	M	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414
4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980

5 rows × 35 columns



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

```
In [16]: ► y_predicted=km.fit_predict(df[["concave points_mean","concave points_worst"]])
y_predicted
```

C:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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(

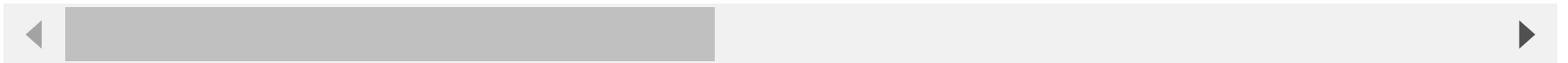
```
Out[16]: array([5, 7, 5, 5, 7, 7, 1, 4, 1, 1, 0, 7, 7, 0, 1, 7, 4, 1, 5, 4, 3, 2,
 5, 1, 1, 5, 5, 4, 1, 4, 7, 4, 7, 7, 7, 7, 4, 2, 6, 1, 0, 4, 5, 4,
 4, 5, 6, 1, 2, 4, 2, 3, 2, 4, 4, 2, 1, 7, 6, 2, 6, 6, 7, 2, 7, 7,
 2, 3, 7, 2, 7, 2, 7, 4, 3, 4, 3, 1, 5, 3, 2, 7, 5, 7, 3, 7, 0, 1,
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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

```
In [17]: df["New Cluster"]=y_predicted  
df.head()
```

Out[17]:

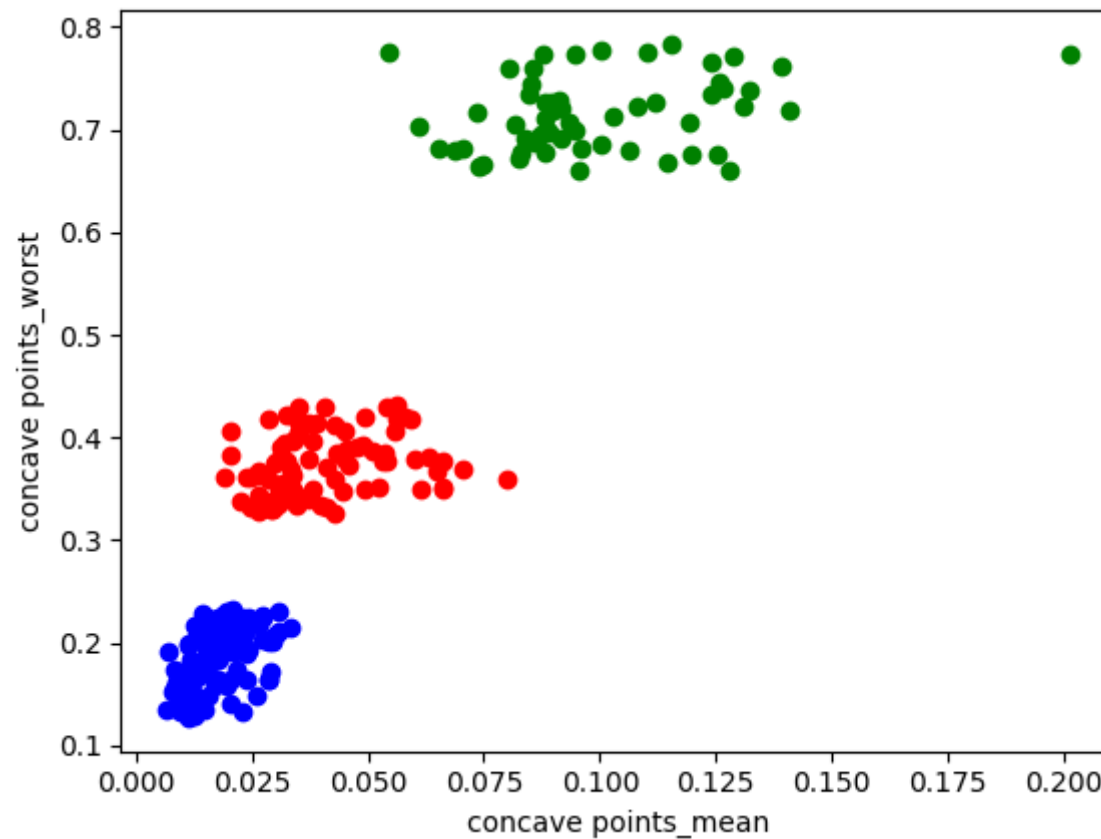
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4	84358402	M	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980


5 rows × 36 columns



```
In [20]: ▶ df1=df[df["New Cluster"]==0]
df2=df[df["New Cluster"]==1]
df3=df[df["New Cluster"]==2]
plt.scatter(df1["concave points_mean"],df1["concave points_worst"],color="red")
plt.scatter(df2["concave points_mean"],df2["concave points_worst"],color="green")
plt.scatter(df3["concave points_mean"],df3["concave points_worst"],color="blue")
plt.xlabel("concave points_mean")
plt.ylabel("concave points_worst")
```

Out[20]: Text(0, 0.5, 'concave points_worst')

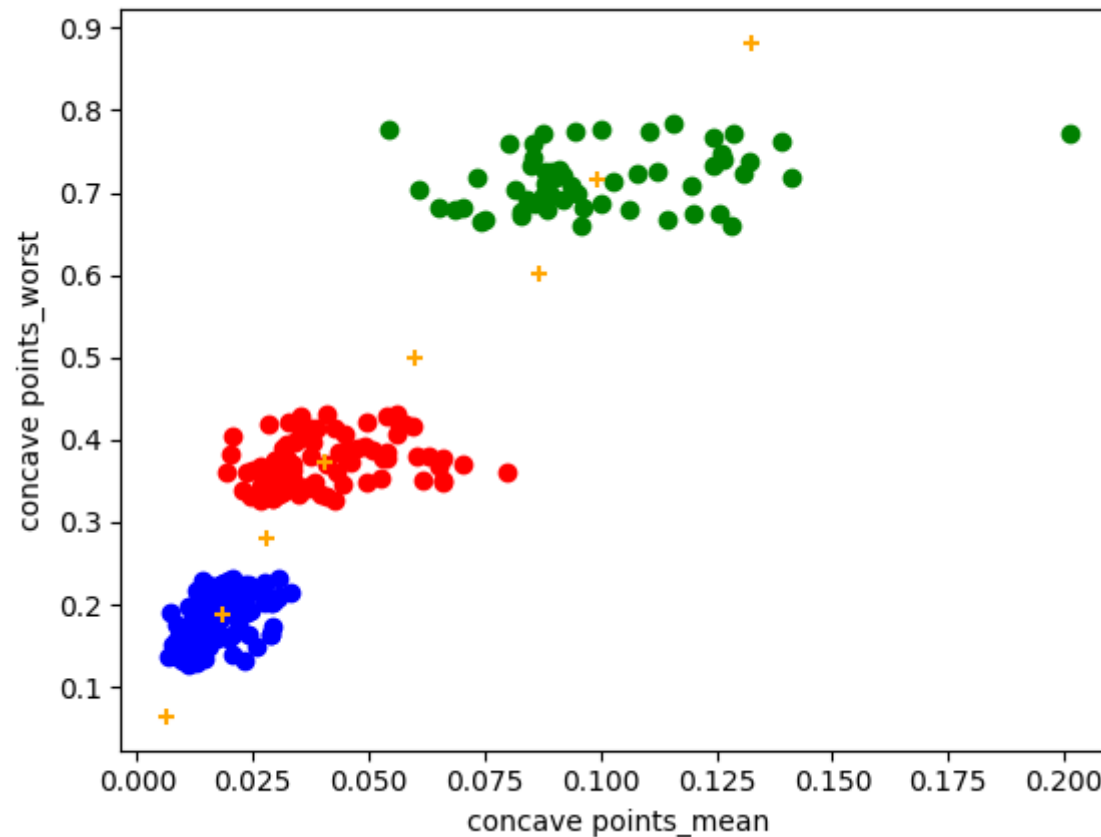


In [21]:  km.cluster_centers_

Out[21]: array([[0.04060904, 0.37325964],
[0.09937034, 0.71544614],
[0.01822572, 0.18850648],
[0.02782514, 0.27945112],
[0.05999863, 0.49852187],
[0.13261516, 0.88198648],
[0.00627369, 0.06437132],
[0.08671579, 0.60248387]])

```
In [22]: ▶ df1=df[df["New Cluster"]==0]
df2=df[df["New Cluster"]==1]
df3=df[df["New Cluster"]==2]
plt.scatter(df1["concave points_mean"],df1["concave points_worst"],color="red")
plt.scatter(df2["concave points_mean"],df2["concave points_worst"],color="green")
plt.scatter(df3["concave points_mean"],df3["concave points_worst"],color="blue")
plt.scatter(km.cluster_centers_[0],km.cluster_centers_[1],color="orange",marker="+")
plt.xlabel("concave points_mean")
plt.ylabel("concave points_worst")
```

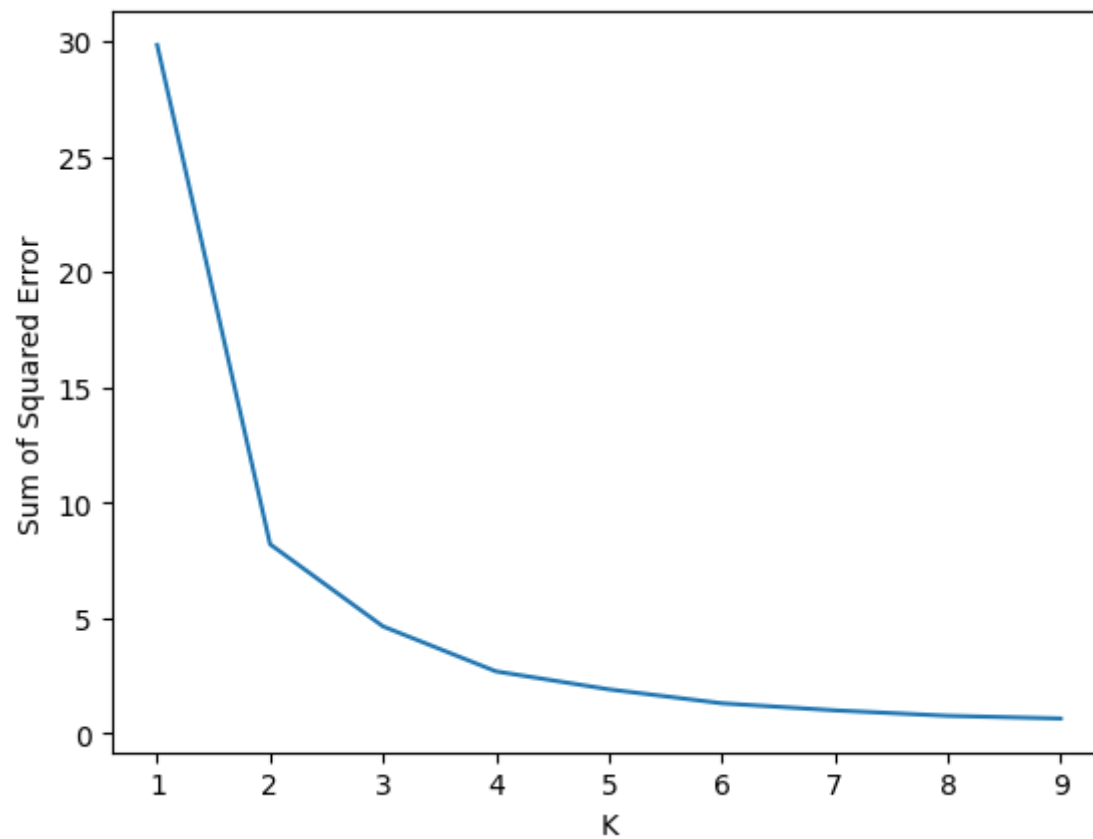
Out[22]: Text(0, 0.5, 'concave points_worst')




```
In [23]: ▶ k_rng=range(1,10)
sse=[]
for k in k_rng:
    km=KMeans(n_clusters=k)
    km.fit(df[["concave points_mean", "concave points_worst"]])
    sse.append(km.inertia_) #km.inertia_ will give you the value of sum of sqa
print(sse)
plt.plot(k_rng,sse)
plt.xlabel("K")
plt.ylabel("Sum of Squared Error")
```

```
C:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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:\Users\MY HOME\AppData\Local\Programs\Python\Python311\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(
[29.83669529147959, 8.192771108043772, 4.632302769625648, 2.682394795326975, 1.902932932070331, 1.2992164832655178, 0.9926335747828674, 0.7560326485435538, 0.6391630303515827]
```

```
Out[23]: Text(0, 0.5, 'Sum of Squared Error')
```



CONCLUSION:

----->for the given dataset we perform kmeans algorithm and we have divided the dataset into several clusters.