

## PROGRAM 8

```
Program 8

import matplotlib.pyplot as plt
from sklearn import datasets
from sklearn.cluster import KMeans
import pandas as pd
import numpy as np

iris = datasets.load_iris()
X = pd.DataFrame(iris.data)
X.columns = ['Sepal Length', 'Sepal Width', 'Petal Length',
             'Petal Width']
y = pd.DataFrame(iris.target)
y.columns = ['Targets']
model = KMeans(n_clusters=3)
model.fit(X)

plt.figure(figsize=(12,12))
colormap = np.array(['red', 'blue', 'black'])
plt.subplot(2,2,1)
plt.scatter(X.Petal.Length, X.Petal.Width, c=colormap[
    y.Targets], s=40)
plt.title('Real Clusters')
plt.xlabel('Petal.Length')
plt.ylabel('Petal.Width')

plt.subplot(2,2,2)
```

```

plt.scatter(X.Petal.Length, X.Petal.Width, c=columns,
            [models.labels], s=40)
plt.title('K-Means Clustering')
plt.xlabel('Petal.Length')
plt.ylabel('Petal.Width')

from sklearn import preprocessing
scaler = preprocessing.StandardScaler()
scaler.fit(X)
xsa = scaler.transform(X)
xs = pd.DataFrame(xsa, columns=X.columns, s)

from sklearn.mixture import
GaussianMixture gmm = GaussianMixture(n_components=3)
gmm.fit(xs)
gmm.y = gmm.predict(xs)
plt.subplot(2, 2, 3)
plt.scatter(X.Petal.Length, X.Petal.Width, c=columns,
            [gmm.y], s=40)
plt.title('GMM Clustering')
plt.xlabel('Petal.Length')
plt.ylabel('Petal.Width')
print('Observation: The GMM using EM algorithm
      based clustering matched the true labels more
      closely than the Kmeans')
plt.show()

```

## OUTPUT

### DATASET 1-hear\_failure

A2- program no 1 x prog8(output).pdf x (12) WhatsApp x MLL/ x 8b - Jupyter Note x 8a - Jupyter Note x Untitled document x + -

localhost:8888/notebooks/MLL/8a.ipynb

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In [1]:

```
import numpy as np
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
from sklearn.mixture import GaussianMixture
import pandas as pd
```

In [2]:

```
X=pd.read_csv("heart_failure.csv")
x1 = X['age'].values
x2 = X['anaemia'].values
X = np.array(list(zip(x1, x2))).reshape(len(x1), 2)
```

In [ ]:

In [ ]:

In [ ]:

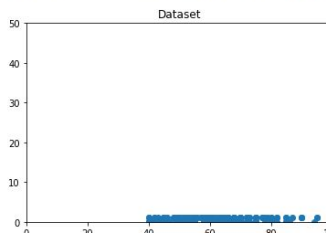
In [ ]:

In [18]:

```
plt.plot()
plt.xlim([0, 100])
plt.ylim([0, 50])
plt.title('Dataset')
plt.scatter(x1, x2)
```

Out[18]: <matplotlib.collections.PathCollection at 0xid3a4b61f08>

Dataset



A2- program no 1 x prog8(output).pdf x (12) WhatsApp x MLL/ x 8b - Jupyter Note x 8a - Jupyter Note x Untitled document x + -

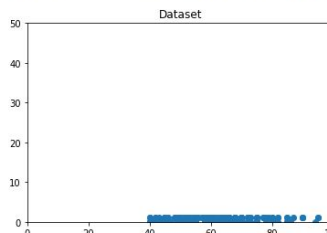
localhost:8888/notebooks/MLL/8a.ipynb

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Out[18]: <matplotlib.collections.PathCollection at 0xid3a4b61f08>

Dataset



In [8]:

```
plt.show()
```

In [9]:

```
gmm = GaussianMixture(n_components=3)
gmm.fit(X)
em_predictions = gmm.predict(X)
```

In [10]:

```
print("\nEM predictions")
print(em_predictions)
print("mean:\n",gmm.means_)
print('\n')
print("Covariances\n",gmm.covariances_)
print(X)
```

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```
print(X)
```

EM predictions

```
[1 0 0 2 0 1 1 0 0 1 1 0 2 2 2 1 1 2 0 2 0 0 0 2 1 1 1 1 0 1 1 1 2 2 0 1 1
1 0 0 1 2 1 1 0 2 2 0 1 0 0 2 0 0 0 1 0 0 2 1 2 2 0 2 2 0 2 1 1 0 2 0 1 0
1 0 1 2 1 0 1 1 0 1 0 2 0 0 2 0 1 0 2 0 0 0 0 0 0 0 0 1 1 2 0 1 0 2 0 2 1
0 2 0 0 0 1 0 1 0 0 0 0 0 2 2 0 0 2 2 0 2 0 1 1 0 0 0 2 1 2 2 0 1 2 2 0
1 0 1 0 2 2 0 0 2 2 1 0 0 2 0 2 2 1 2 0 0 1 2 2 0 2 0 0 1 2 0 0 2 0 0 1 0
0 2 0 0 2 1 0 2 1 2 1 2 0 2 0 0 2 1 0 1 2 2 1 0 2 1 2 1 2 0 1 1 2 0 0 1 0
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2 0 0 0 0 0 2 0 0 2 2 2 2 1 2 0 2 1 1 2 0 1 1 2 0 2 0 0 2 0 1 2 0 2 0 0 0
2 2 2]
```

mean:

```
[[61.44675838  0.58816321]
 [73.41882093  0.30783914]
 [48.19847624  0.36707877]]
```

Covariances

```
[[[24.5074528  0.25456207]
 [ 0.25456207  0.24222825]]
 [[81.97328576  1.58440483]
 [ 1.58440483  0.21307521]]
 [[21.69883636  0.36083588]
 [ 0.36083588  0.23233295]]]
```

In [11]:

```
plt.title('Exception Maximum')
plt.scatter(X[:,0], X[:,1],c=em_predictions,s=50)
plt.show()
```

Exception Maximum

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```
print(X)
```

Covariances

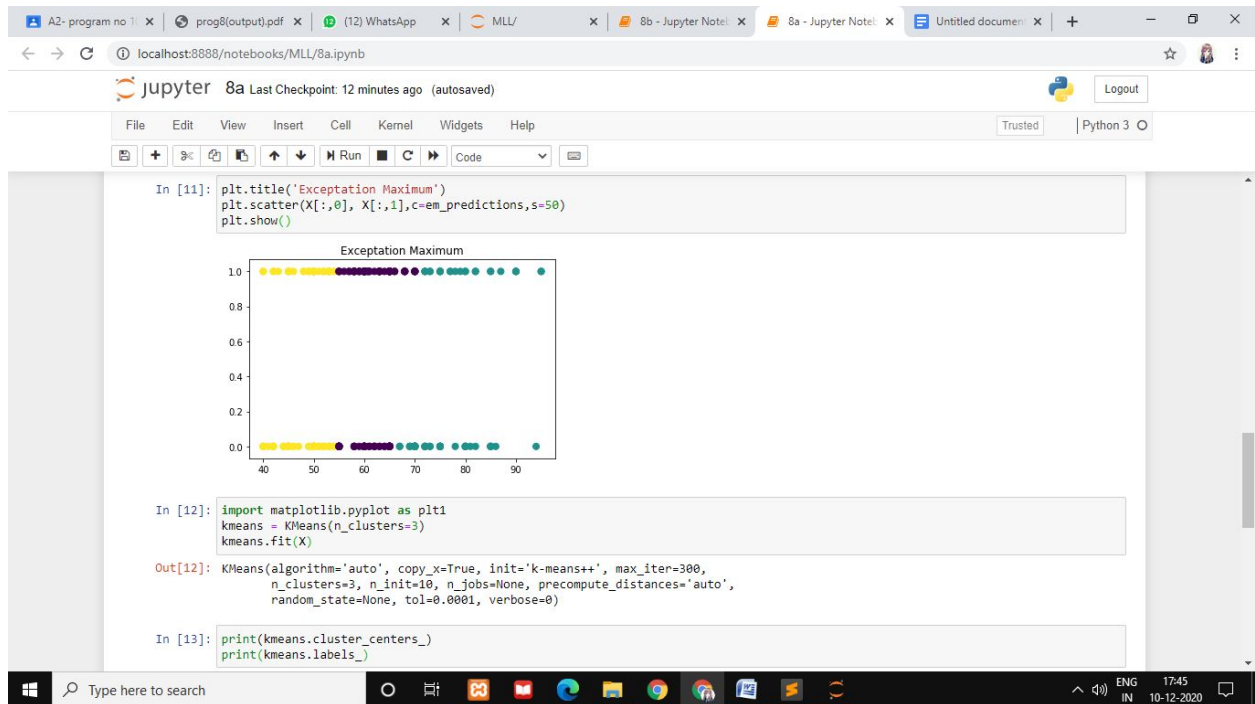
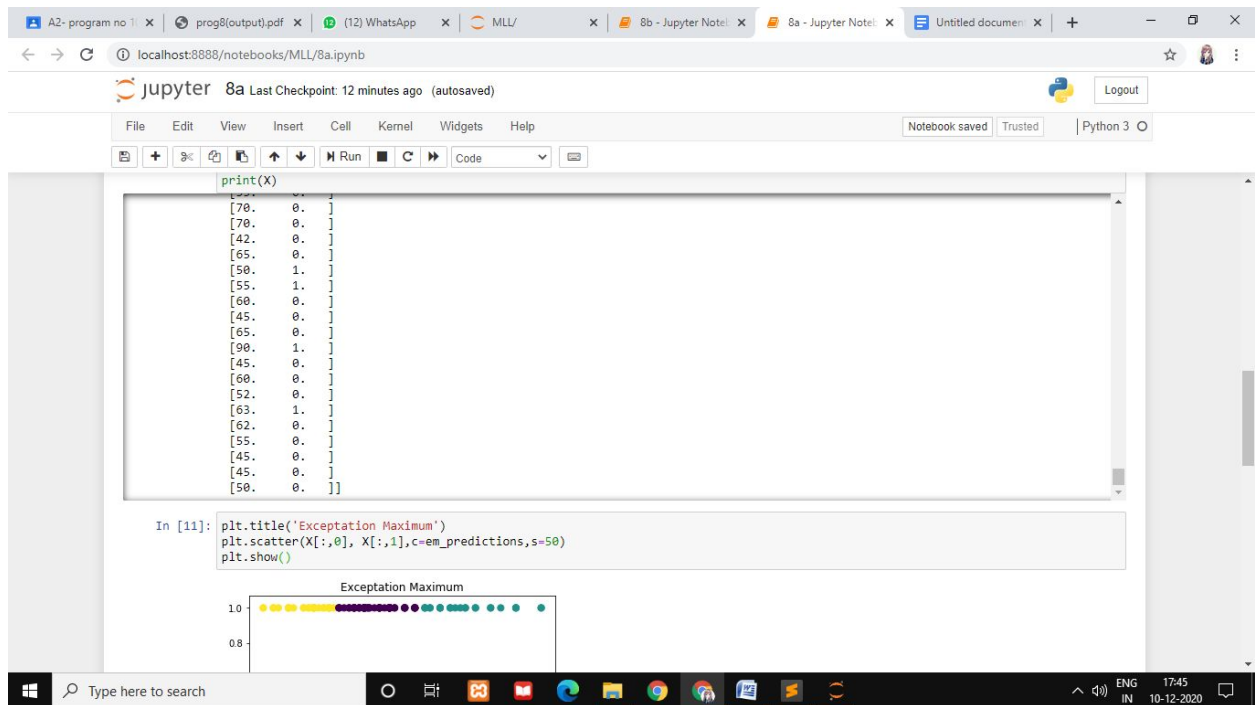
```
[[[24.5074528  0.25456207]
 [ 0.25456207  0.24222825]]
 [[81.97328576  1.58440483]
 [ 1.58440483  0.21307521]]
 [[21.69883636  0.36083588]
 [ 0.36083588  0.23233295]]]
```

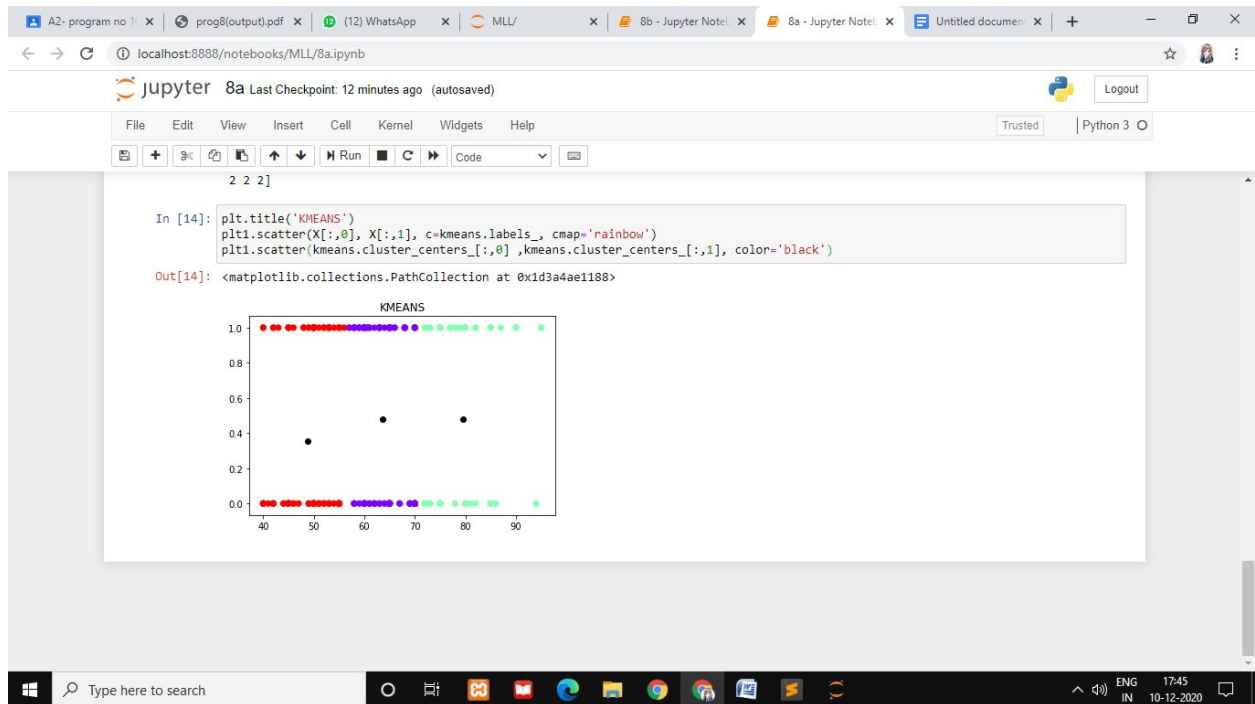
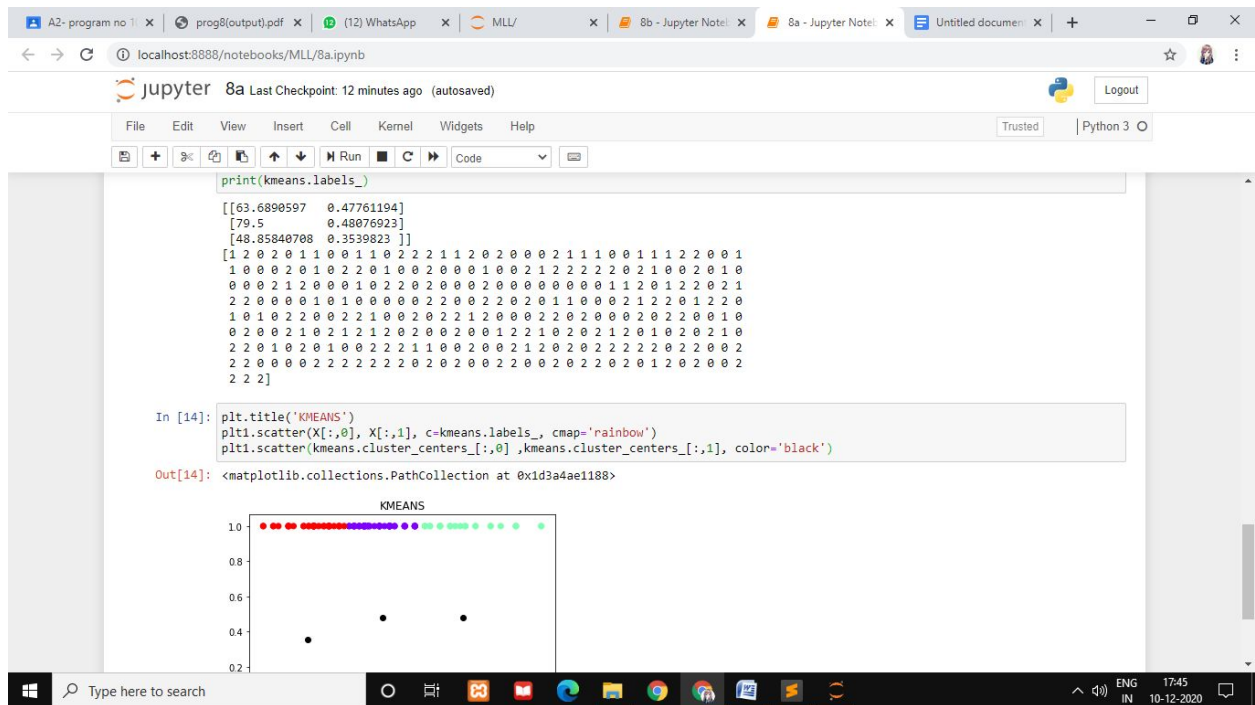
In [11]:

```
plt.title('Exception Maximum')
plt.scatter(X[:,0], X[:,1],c=em_predictions,s=50)
plt.show()
```

Exception Maximum







## DATASET 2- Iris Dataset

localhost:8888/notebooks/MLL/8b.ipynb

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```
In [1]: import matplotlib.pyplot as plt
from sklearn import datasets
from sklearn.cluster import KMeans
import pandas as pd
import numpy as np

In [2]: iris = datasets.load_iris()
X = pd.DataFrame(iris.data)
X.columns = ['Sepal_Length', 'Sepal_Width', 'Petal_Length', 'Petal_Width']
y = pd.DataFrame(iris.target)
y.columns = ['Targets']

In [3]: model = KMeans(n_clusters=3)
model.fit(X)

Out[3]: KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,
n_clusters=3, n_init=10, n_jobs=None, precompute_distances='auto',
random_state=None, tol=0.0001, verbose=0)

In [4]: plt.figure(figsize=(12,12))
colormap = np.array(['red', 'lime', 'black'])

<Figure size 864x864 with 0 Axes>

In [5]: plt.subplot(2, 2, 1)
plt.scatter(X.Petal_Length, X.Petal_Width, c=colormap[y.Targets], s=40)
plt.title('Real Clusters')
plt.xlabel('Petal_Length')
```

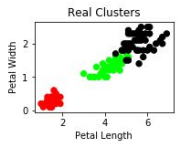
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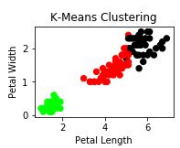
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```
Out[5]: Text(0, 0.5, 'Petal Width')
```



```
In [6]: plt.subplot(2, 2, 2)
plt.scatter(X.Petal_Length, X.Petal_Width, c=colormap[model.labels_], s=40)
plt.title('K-Means Clustering')
plt.xlabel('Petal Length')
plt.ylabel('Petal Width')
```

```
Out[6]: Text(0, 0.5, 'Petal Width')
```



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```
In [7]: from sklearn import preprocessing
scaler=preprocessing.StandardScaler()
scaler.fit(X)

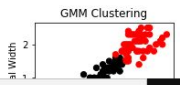
Out[7]: StandardScaler(copy=True, with_mean=True, with_std=True)

In [8]: xsa=scaler.transform(X)
xs=pd.DataFrame(xsa,columns=X.columns)
from sklearn.mixture import GaussianMixture
gmm=GaussianMixture(n_components=3)
gmm.fit(xs)

Out[8]: GaussianMixture(covariance_type='full', init_params='kmeans', max_iter=100,
means_init=None, n_components=3, n_init=1, precisions_init=None,
random_state=None, reg_covar=1e-06, tol=0.001, verbose=0,
verbose_interval=10, warm_start=False, weights_init=None)

In [9]: gmm_y=gmm.predict(xs)
plt.subplot(2,2,3)
plt.scatter(X.Petal_Length,X.Petal_Width,c=colormap[gmm_y],s=40)
plt.title('GMM Clustering')
plt.xlabel('Petal Length')
plt.ylabel('Petal Width')

Out[9]: Text(0, 0.5, 'Petal Width')
```



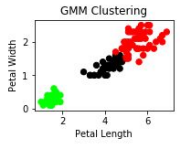
localhost:8888/notebooks/MLL/8b.ipynb

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```
plt.scatter(X.Petal_Length,X.Petal_Width,c=colormap[gmm_y],s=40)
plt.title('GMM Clustering')
plt.xlabel('Petal Length')
plt.ylabel('Petal Width')

Out[9]: Text(0, 0.5, 'Petal Width')
```



```
In [10]: print('Observation: The GMM using EM algorithm based clustering matched the true labels more closely than the Kmeans')
plt.show()

Observation: The GMM using EM algorithm based clustering matched the true labels more closely than the Kmeans

In [ ]:
```