

# Machine Learning Lab Assignment 3

Name – Rishav Kundu

Roll – 001811001039

Semester – 7

## **PART 1**

### **1) Wine Dataset**

#### **1.1) GaussianHMM Without Tuning**



Confusion Matrix:

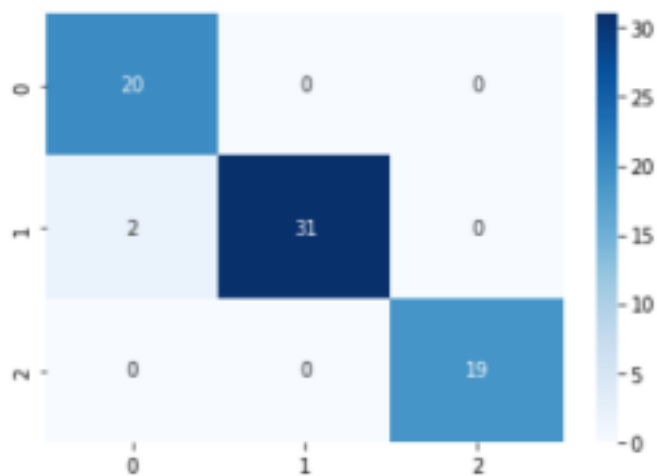
```
[[20  0  0]
 [ 2 31  0]
 [ 0  0 19]]
```

Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 1            | 0.91      | 1.00   | 0.95     | 20      |
| 2            | 1.00      | 0.94   | 0.97     | 33      |
| 3            | 1.00      | 1.00   | 1.00     | 19      |
| accuracy     |           |        | 0.97     | 72      |
| macro avg    | 0.97      | 0.98   | 0.97     | 72      |
| weighted avg | 0.97      | 0.97   | 0.97     | 72      |

Accuracy:

0.9722222222222222



## 1.2) GaussianHMM With Tuning

Confusion Matrix:

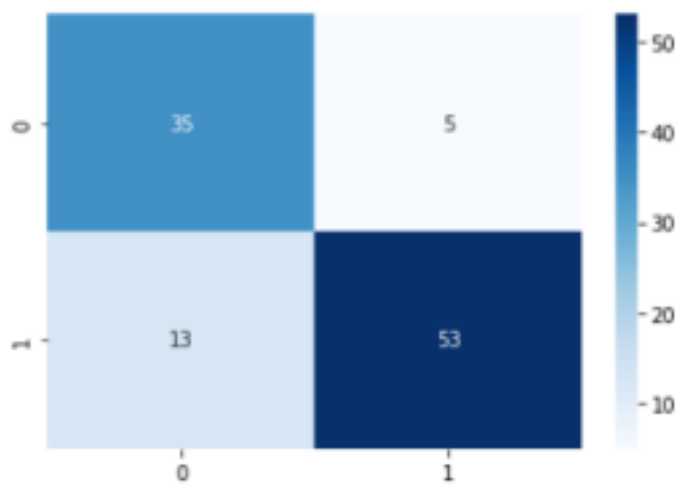
```
[[35  5]
 [13 53]]
```

Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

Accuracy:

0.8301886792452831



### 1.3) GMMHMM Without Tuning



Confusion Matrix:



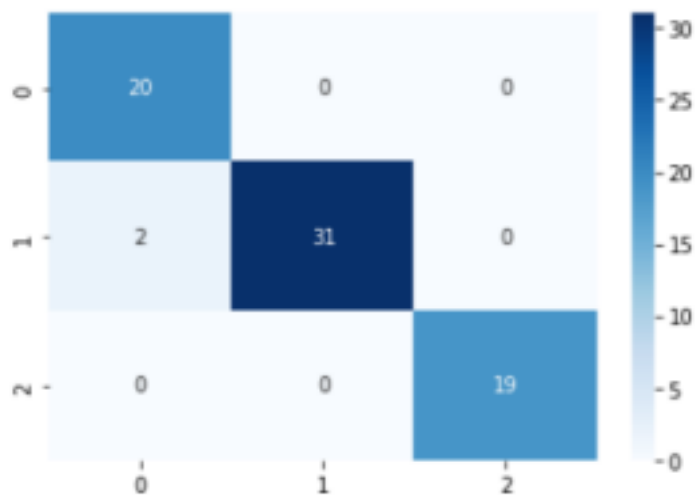
```
[[20  0  0]
 [ 2 31  0]
 [ 0  0 19]]
```

Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 1            | 0.91      | 1.00   | 0.95     | 20      |
| 2            | 1.00      | 0.94   | 0.97     | 33      |
| 3            | 1.00      | 1.00   | 1.00     | 19      |
| accuracy     |           |        | 0.97     | 72      |
| macro avg    | 0.97      | 0.98   | 0.97     | 72      |
| weighted avg | 0.97      | 0.97   | 0.97     | 72      |

Accuracy:

0.9722222222222222



## 1.4) GMMHMM With Tuning

➤ Confusion Matrix:

```
[[35  5]
 [13 53]]
```

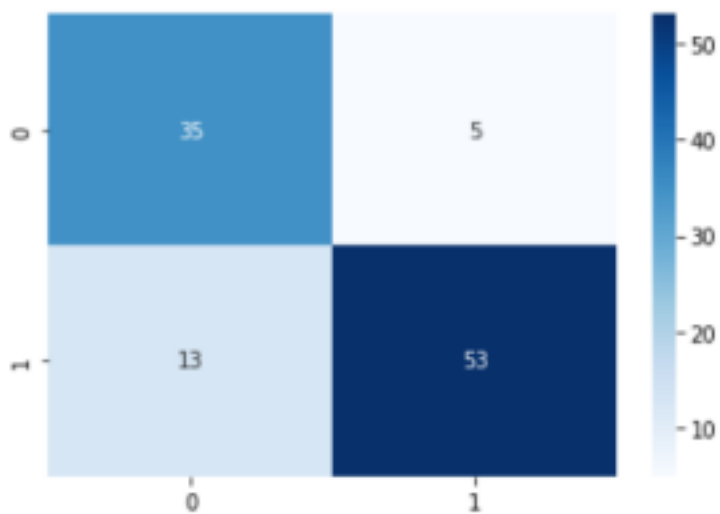
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----

Accuracy:

0.8301886792452831



## 1.5) MultinomialHMM Without Tuning

↳ Confusion Matrix:

```
[[35  5]
 [13 53]]
```

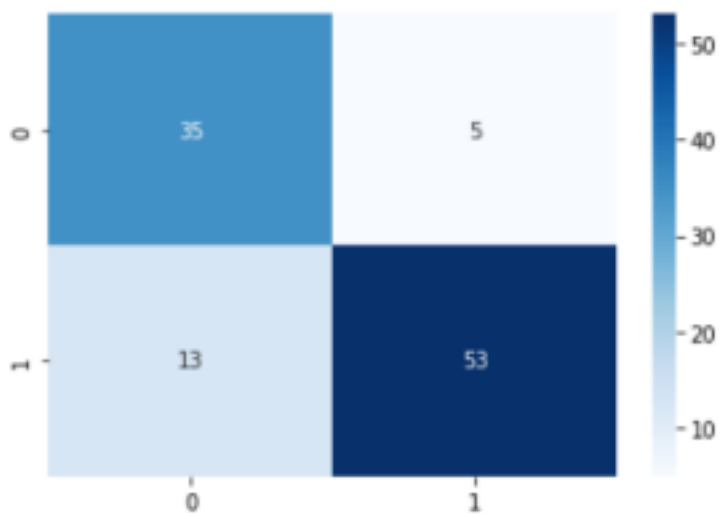
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----

Accuracy:

0.8301886792452831



## 1.6) MultinomialHMM Without Tuning

Confusion Matrix:

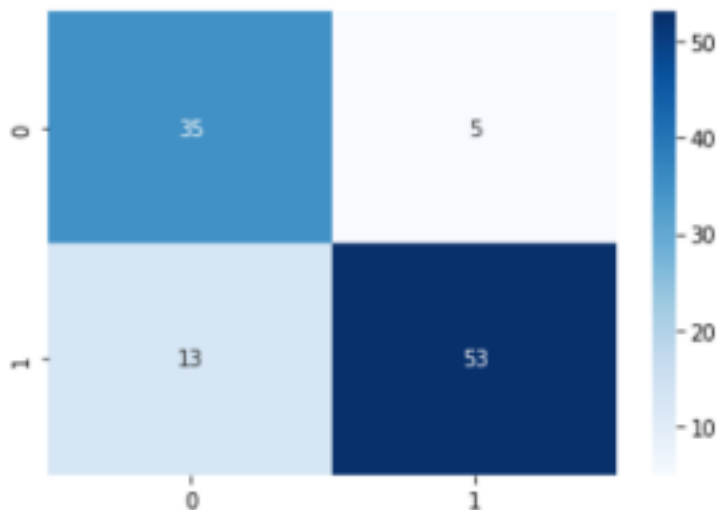
```
[[35  5]
 [13 53]]
```

Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

Accuracy:

0.8301886792452831



The maximum accuracy was achieved when the Train-Test split ratio was 70:30, which was achieved by using the Gaussian Model. The maximum range of accuracies was achieved by the Gaussian Model, followed by the GMMHMM model, which is followed by the MultinomialHMM model.

## 2) Ionosphere Dataset

## 2.1) GaussianHMM Without Tuning

```
➤ Confusion Matrix:
[[35  5]
 [13 53]]

-----

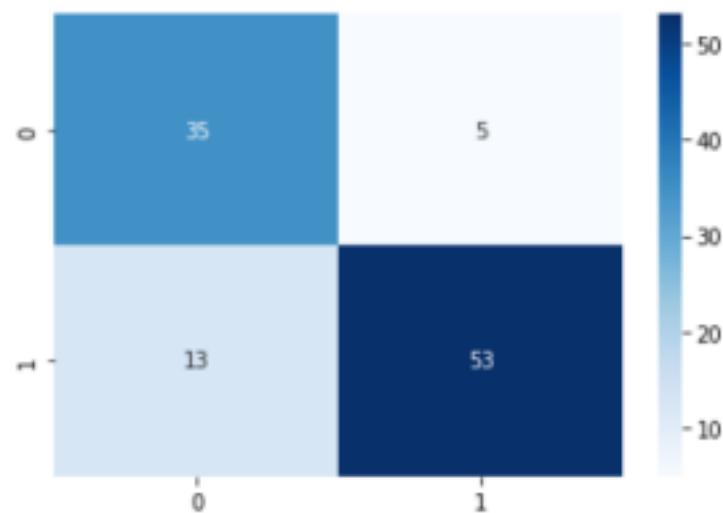
Performance Evaluation
      precision    recall  f1-score   support

      b         0.73      0.88      0.80         40
      g         0.91      0.80      0.85         66

 accuracy              0.83         106
macro avg              0.82      0.84      0.83         106
weighted avg          0.84      0.83      0.83         106

-----

Accuracy:
0.8301886792452831
```



## 2.2) GaussianHMM With Tuning



➤ Confusion Matrix:

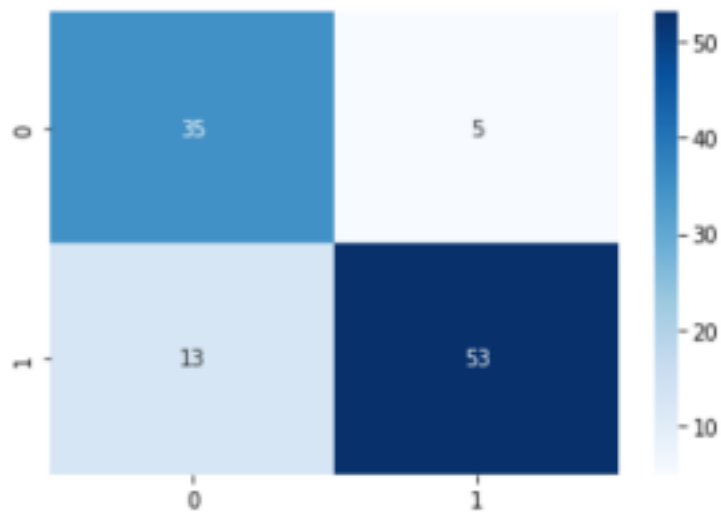
```
[[35  5]
 [13 53]]
```

-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----  
Accuracy:

0.8301886792452831



## 2.3) GMMHMM Without Tuning

➤ Confusion Matrix:

```
[[35  5]
 [13 53]]
```

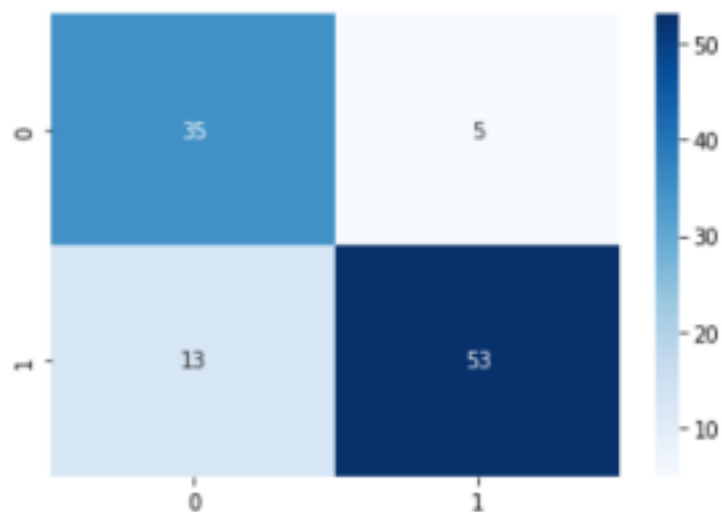
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----

Accuracy:

0.8301886792452831



## 2.4) GMMHMM With Tuning

➤ Confusion Matrix:

```
[[35  5]
 [13 53]]
```

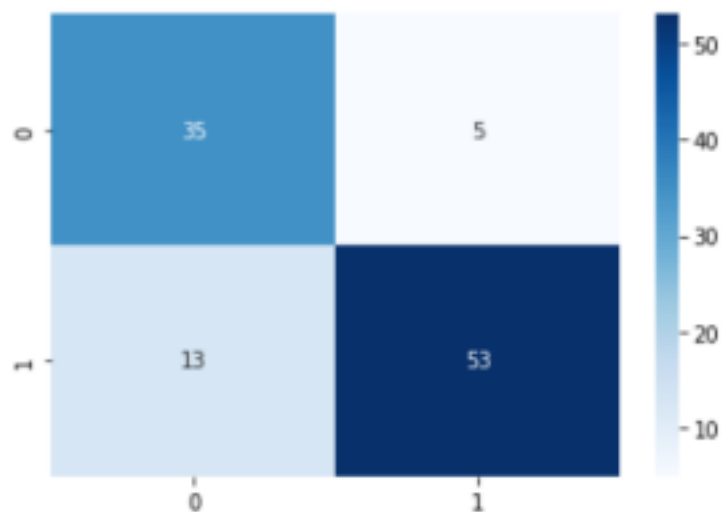
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----

Accuracy:

0.8301886792452831



## 2.5) MultinomialHMM Without Tuning

➤ Confusion Matrix:

```
[[35  5]
 [13 53]]
```

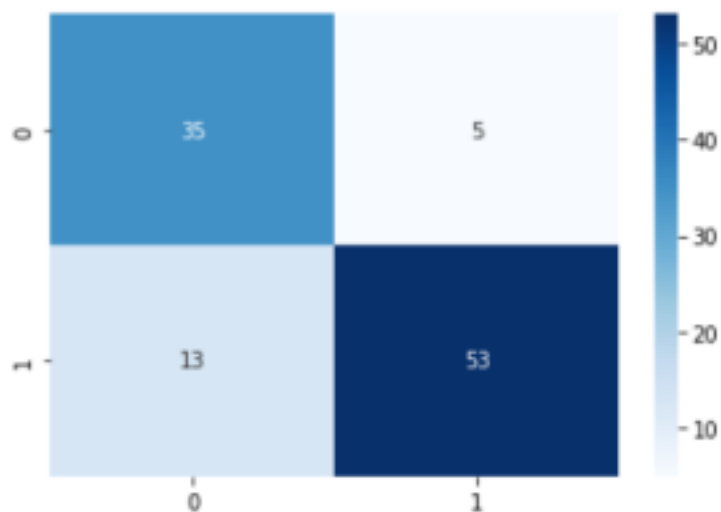
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| b            | 0.73      | 0.88   | 0.80     | 40      |
| g            | 0.91      | 0.80   | 0.85     | 66      |
| accuracy     |           |        | 0.83     | 106     |
| macro avg    | 0.82      | 0.84   | 0.83     | 106     |
| weighted avg | 0.84      | 0.83   | 0.83     | 106     |

-----  
-----

Accuracy:

0.8301886792452831



## 2.6) MultinomialHMM Without Tuning

The maximum accuracy was achieved when the Train-Test split ratio was 70:30, which was achieved by using the Gaussian Model. The maximum range of accuracies was achieved by the Gaussian Model, followed by the GMMHMM model, which is

followed by the MultinomialHMM model.

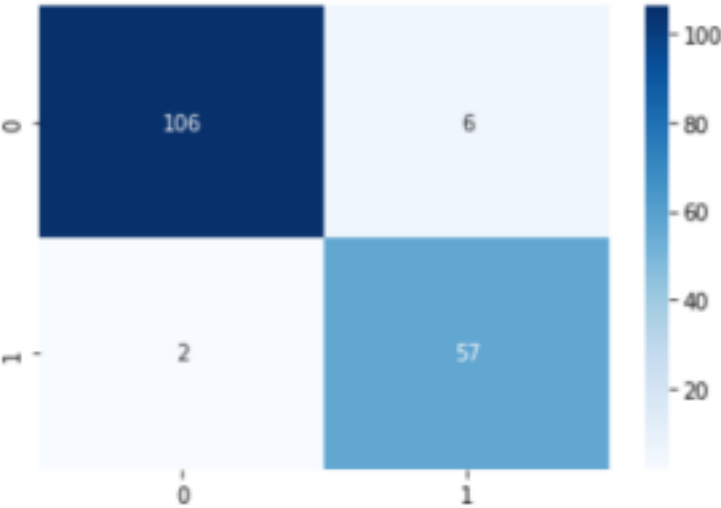
### 3) Breast Cancer Dataset

#### 3.1) GaussianHMM Without Tuning

```
Confusion Matrix:  
[[106   6]  
 [  2  57]]
```

| Performance Evaluation |           |        |          |         |  |
|------------------------|-----------|--------|----------|---------|--|
|                        | precision | recall | f1-score | support |  |
| B                      | 0.98      | 0.95   | 0.96     | 112     |  |
| M                      | 0.90      | 0.97   | 0.93     | 59      |  |
| accuracy               |           |        | 0.95     | 171     |  |
| macro avg              | 0.94      | 0.96   | 0.95     | 171     |  |
| weighted avg           | 0.96      | 0.95   | 0.95     | 171     |  |

Accuracy:  
0.9532163742690059



#### 3.2) GaussianHMM With Tuning

Confusion Matrix:

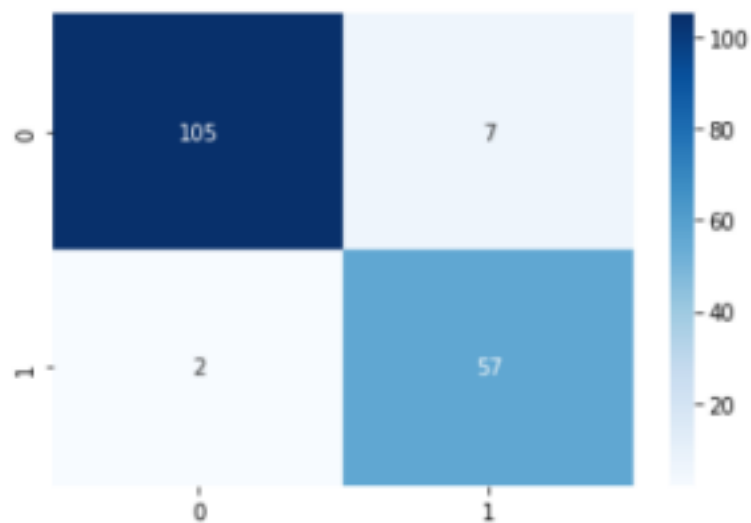
```
[[105  7]
 [  2 57]]
```

Performance Evaluation

|  |              | precision | recall | f1-score | support |
|--|--------------|-----------|--------|----------|---------|
|  | B            | 0.98      | 0.94   | 0.96     | 112     |
|  | M            | 0.89      | 0.97   | 0.93     | 59      |
|  | accuracy     |           |        | 0.95     | 171     |
|  | macro avg    | 0.94      | 0.95   | 0.94     | 171     |
|  | weighted avg | 0.95      | 0.95   | 0.95     | 171     |

Accuracy:

0.9473684210526315



### 3.3) GMMHMM Without Tuning

➤ Confusion Matrix:

```
[[105  7]
 [  5 54]]
```

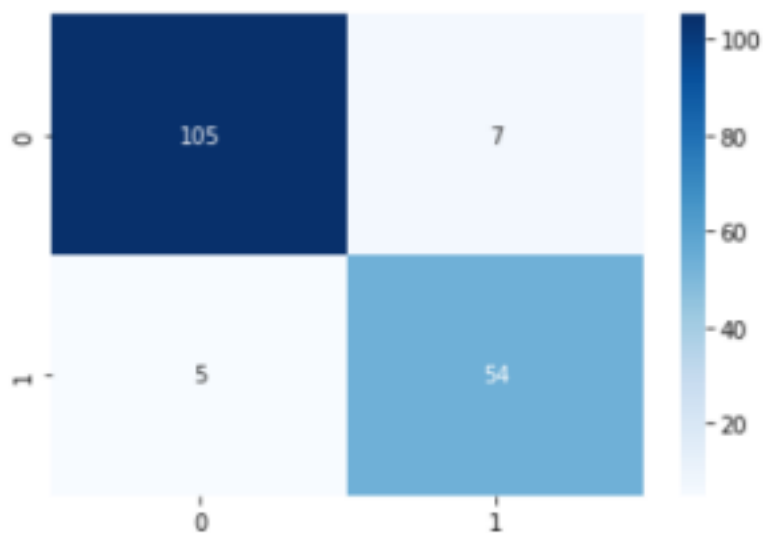
-----  
-----  
Performance Evaluation

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| B            | 0.95      | 0.94   | 0.95     | 112     |
| M            | 0.89      | 0.92   | 0.90     | 59      |
| accuracy     |           |        | 0.93     | 171     |
| macro avg    | 0.92      | 0.93   | 0.92     | 171     |
| weighted avg | 0.93      | 0.93   | 0.93     | 171     |

-----  
-----

Accuracy:

0.9298245614035088



### 3.4) GMMHMM With Tuning



### 3.5) MultinomialHMM Without Tuning





### 3.6) MultinomialHMM Without Tuning



**The maximum accuracy was achieved when the Train-Test split ratio was 70:30, which was achieved by using the Gaussian Model. The maximum range of accuracies was achieved by the Gaussian Model, followed by the GMMHMM model, which is followed by the MultinomialHMM model.**

## **PART 2**

## 1) CIFAR-10





**2) MNIST**



**3) SAVEE**





#### 4) EmoDB





**It was observed that the more layers we add the higher accuracy we can achieve. At the same time, if we keep on adding more layers, the final accuracy will saturate. Also, the number of convolution and the pooling layers play an important role in training the model.**

## **PART 3**

### **1) VGG-16**

#### **1.1) CIFAR-10**



## 1.2) MNIST



## 1.3) SAVEE





#### **1.4) EmoDB**



**The entire model can be broken down into 5 blocks, where each block contains 3 convolution and 1 max-pooling layers.**

Looking at the complexity of the model and the limitations of google colab, I have reduced the input size for the model,i.e., i have taken 2000 training data points and 2000 testing data points.

## **2) ResNet-50**

### **2.1) CIFAR-10**



### **2.2) MNIST**



### 2.3) SAVEE



### 2.4) EmoDB



**Looking at the complexity of the model and the limitations of google colab, I have reduced the input size for the model,i.e., I have taken 2000 training data points and 2000 testing data points.**

## **3) Recurrent Neural Networks (RNN)**

### **3.1) CIFAR-10**



### 3.2) MNIST



### 3.3) SAVEE



### 3.4) EmoDB



**Looking at the complexity of the model and the limitations of google colab, I have reduced the input size for the model,i.e., I have taken 2000 training data points and 2000 testing data points.**

## **4) AlexNet**

### **4.1) CIFAR-10**



### **4.2) MNIST**



#### **4.3) SAVEE**



#### **4.4) EmoDB**



**Looking at the complexity of the model and the limitations of google colab, I have reduced the input size for the model,i.e., I have taken 2000 training data points and 2000 testing data points.**

## **5) GoogleNet**

### **5.1) CIFAR-10**



### **5.2) MNIST**





**5.3) SAVEE**



#### **5.4) EmoDB**

**Looking at the complexity of the model and the limitations of google colab, I have reduced the input size for the model ,i.e., I have taken 2000 training data points and 2000 testing data points.**