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**Subject :- Intro to NLP**

## **Assignment 2:- POS Tagging**

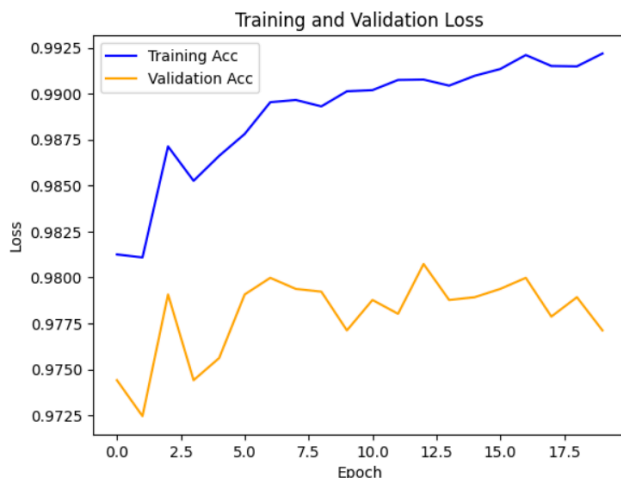
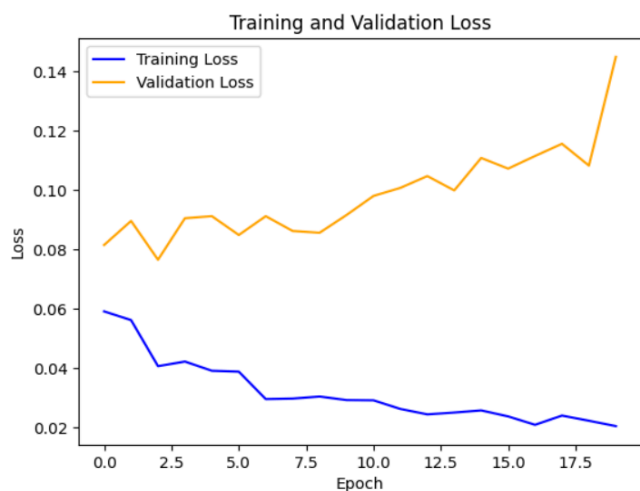
### **❖ Hyperparameter tuning result:-**

#### **1. FFNN Result**

##### **● Configuration 1:-**

```
p = 3
s = 2
BATCH_SIZE = 64
EPOCHS = 10
embedding_dim = 100
HIDDEN_LAYER_DIMENSION = 200

def forward(self, x):
    x = x.to(self.l1.weight.dtype)
    out = self.l1(x)
    out = self.relu(out)
    out = self.l2(out)
    out = self.relu(out)
    out = self.l3(out)
    return out
```



## For Validation:-

===== Validation Data =====

Validation Loss: 0.1448

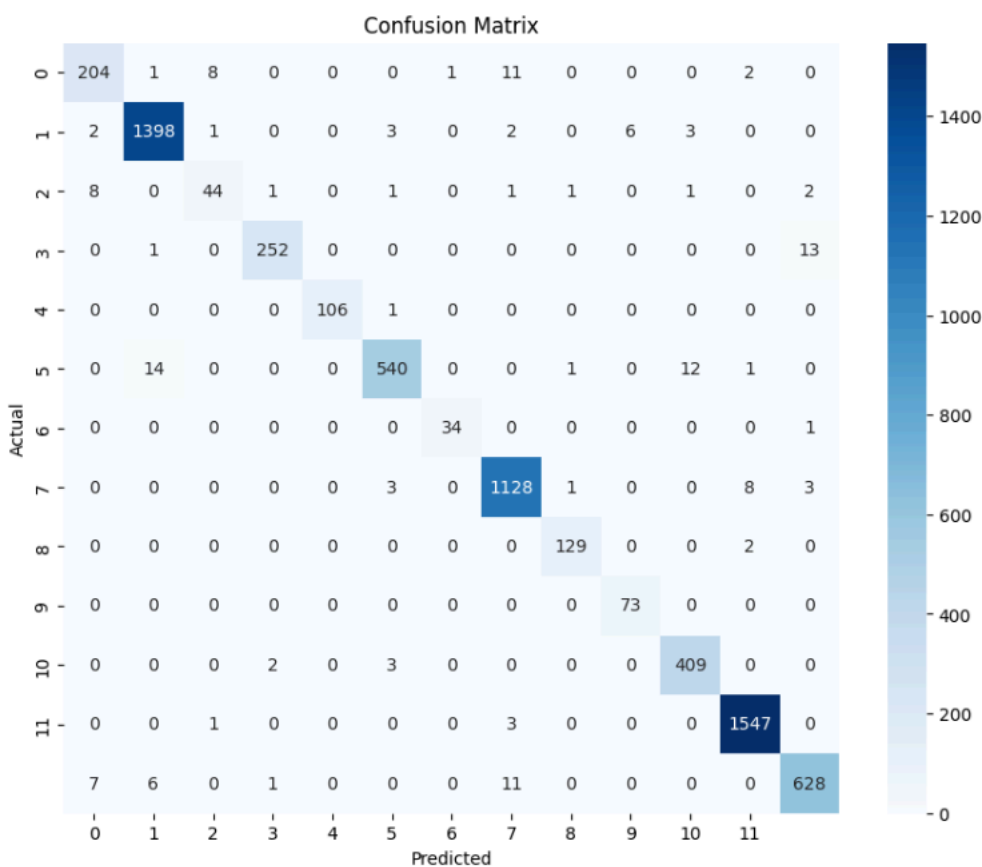
Validation Accuracy: 0.9771

Validation Precision: 0.9993

Validation Recall: 0.9882

Validation F1 Score: 0.9937

=====



## For Test:-

===== Test Data =====

Test Loss: 0.1043

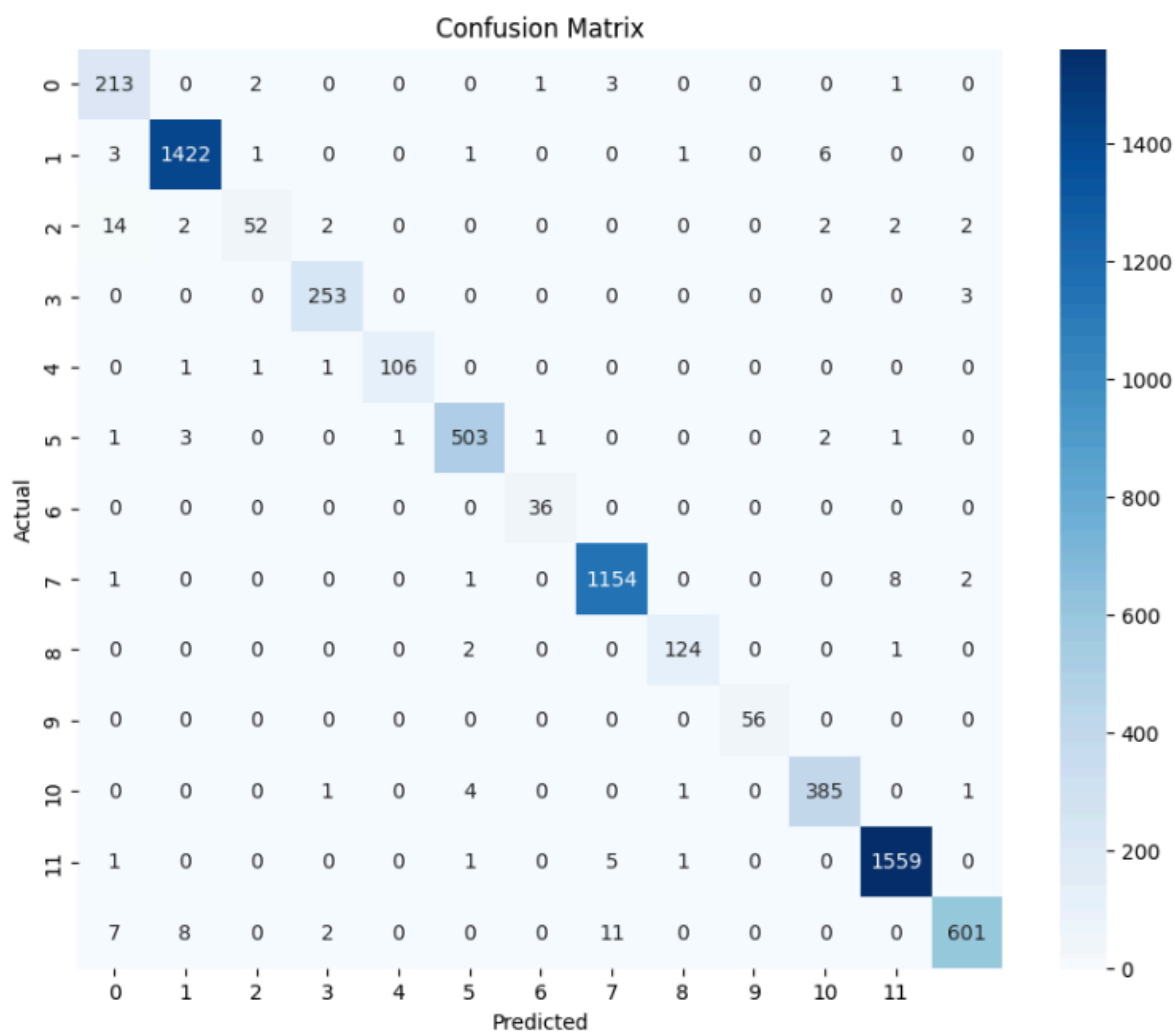
Test Accuracy: 0.9824

Test Precision: 1.0000

Test Recall: 0.9815

Test F1 Score: 0.9907

=====

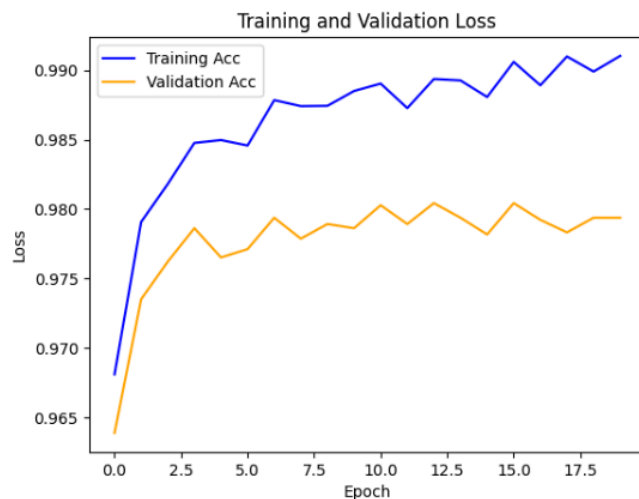


## ● Configuration 2:-

```
p = 2
s = 3
BATCH_SIZE = 64
EPOCHS = 20
embedding_dim = 100
HIDDEN_LAYER_DIMENSION = 100

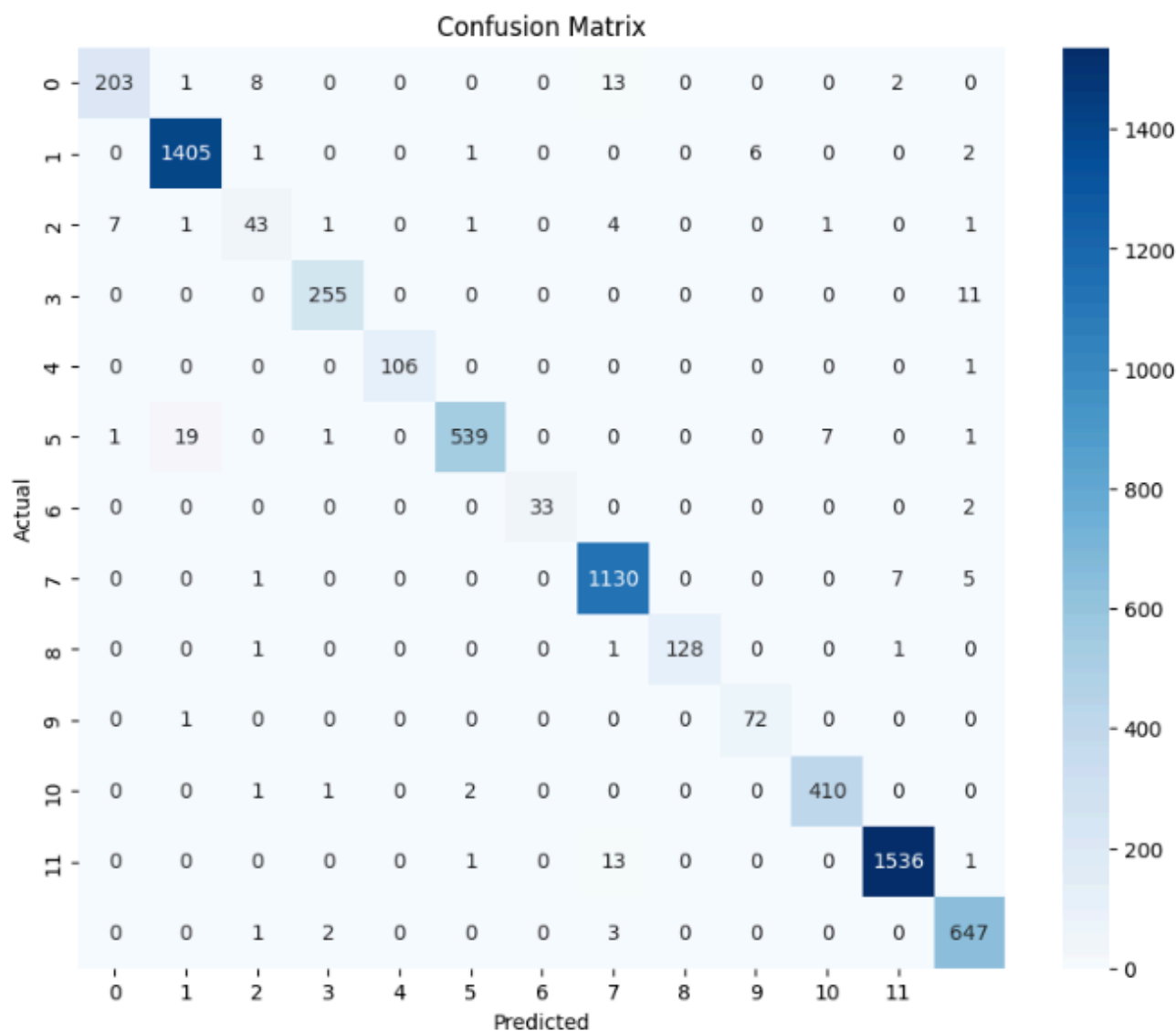
def forward(self, x):
    x = x.to(self.l1.weight.dtype)
    out = self.l1(x)
    out = self.relu(out)
    out = self.l2(out)
    out = self.relu(out)
    out = self.l3(out)
    out = self.relu(out)
    out = self.l4(out)
    out = self.relu(out)
    out = self.l5(out)

    return out
```



**For Validation Data:-**

```
===== Validation Data =====  
Validation Loss:  0.0993  
Validation Accuracy: 0.9794  
Validation Precision: 0.9993  
Validation Recall: 0.9944  
Validation F1 Score: 0.9969  
=====
```



## For Test Data:-

===== Test Data =====

Test Loss: 0.0858

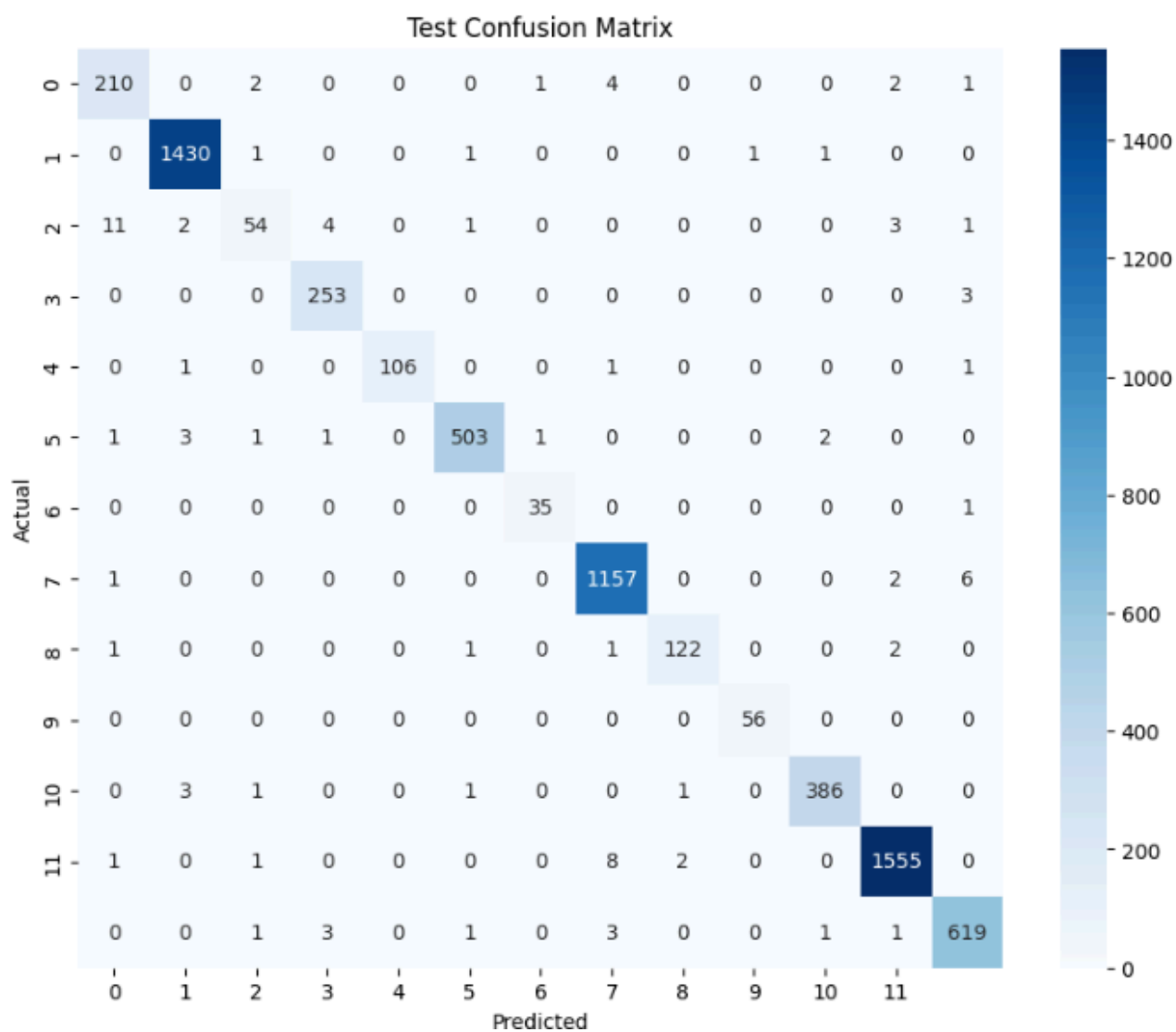
Test Accuracy: 0.9857

Test Precision: 1.0000

Test Recall: 0.9897

Test F1 Score: 0.9948

=====

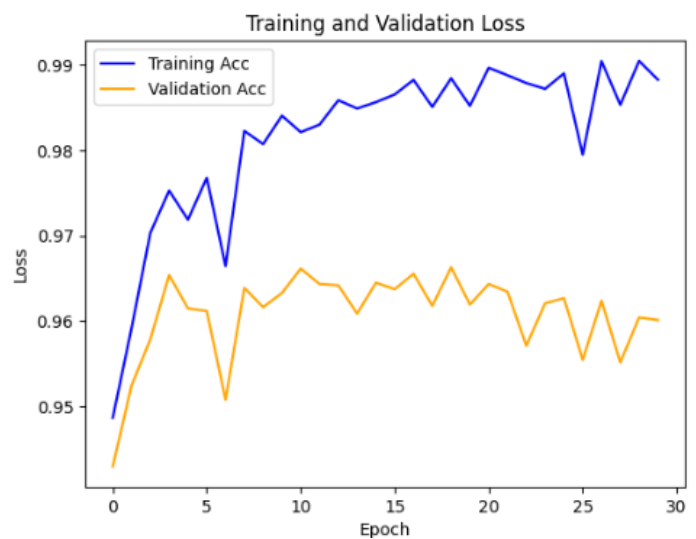
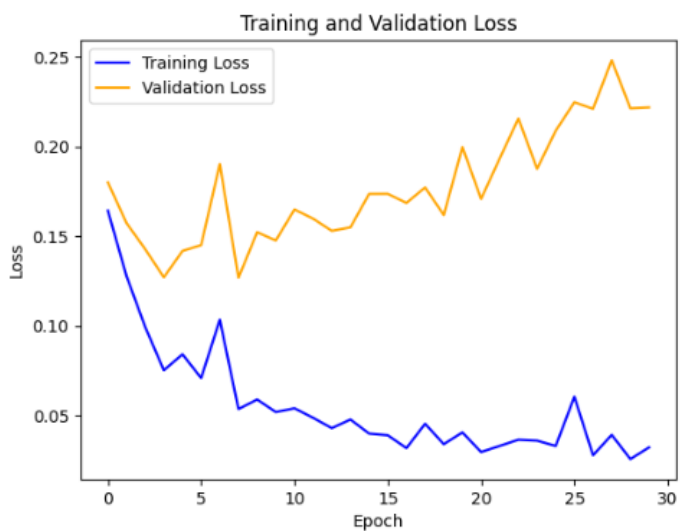


## ● Configuration 3:-

```
p = 4
s = 5
BATCH_SIZE = 64
round_precision = 5
EPOCHS = 30
embedding_dim = 100
HIDDEN_LAYER_DIMENSION = 200

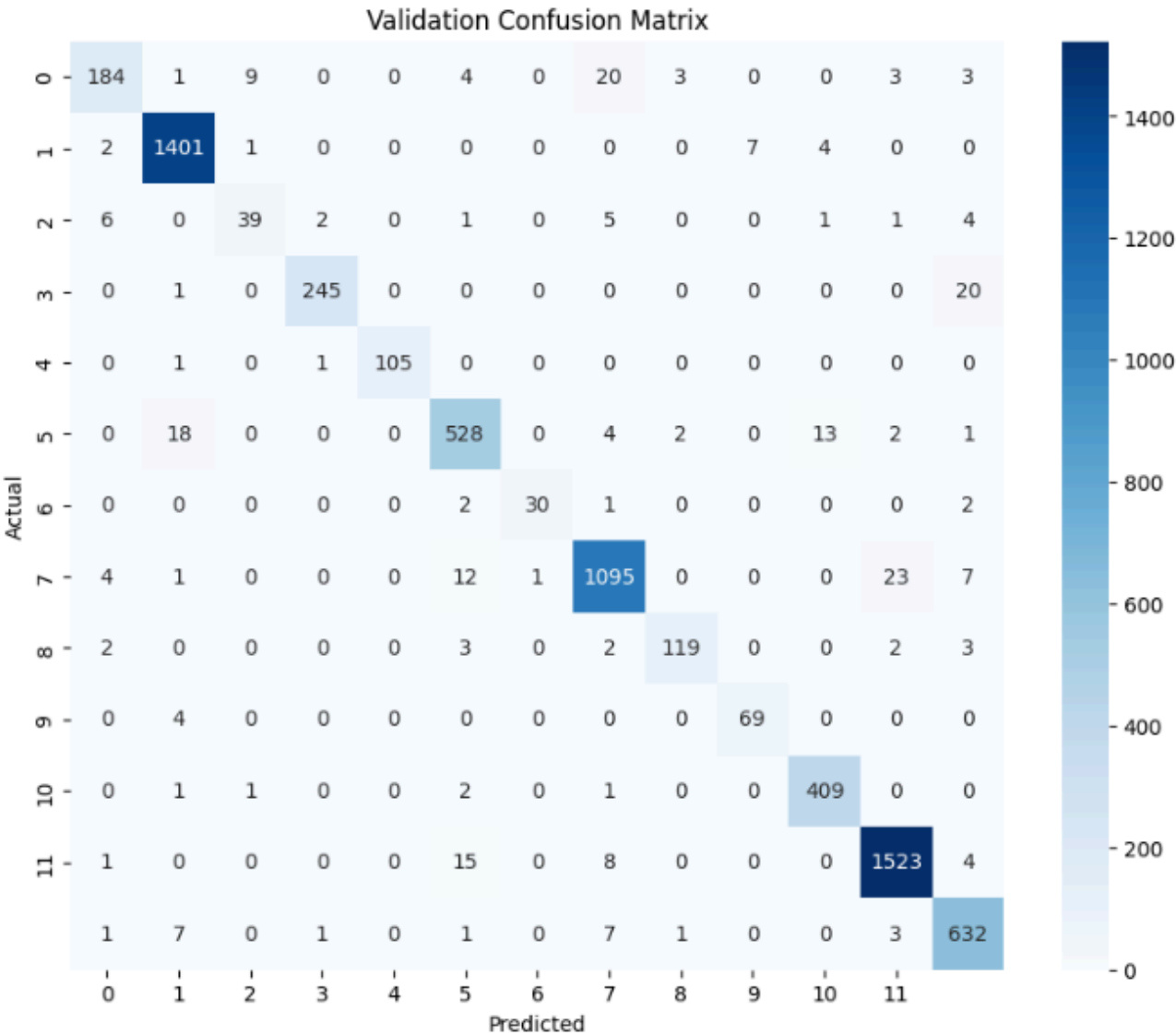
def forward(self, x):
    x = x.to(self.l1.weight.dtype)
    out = self.l1(x)
    out = self.leaky_relu(out)
    out = self.l2(out)
    out = self.leaky_relu(out)
    out = self.l3(out)
    out = self.leaky_relu(out)
    out = self.l4(out)

    return out
```



For Validation Data:-

===== Validation Data =====  
Validation Loss: 0.2217  
Validation Accuracy: 0.9601  
Validation Precision: 0.9993  
Validation Recall: 0.9890  
Validation F1 Score: 0.9941  
=====





## For test data:-

===== Test Data =====

Test Loss: 0.2062

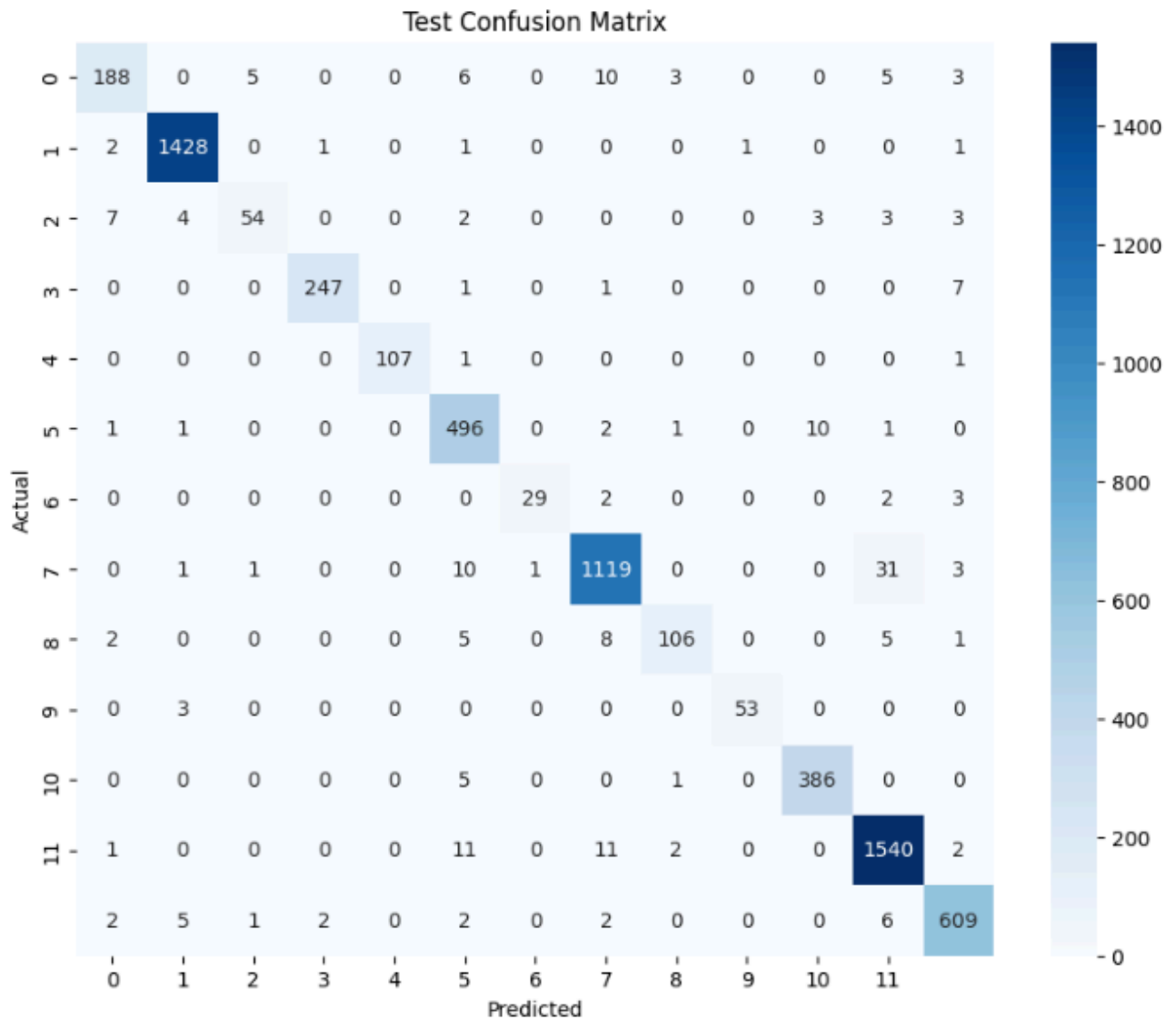
Test Accuracy: 0.9669

Test Precision: 1.0000

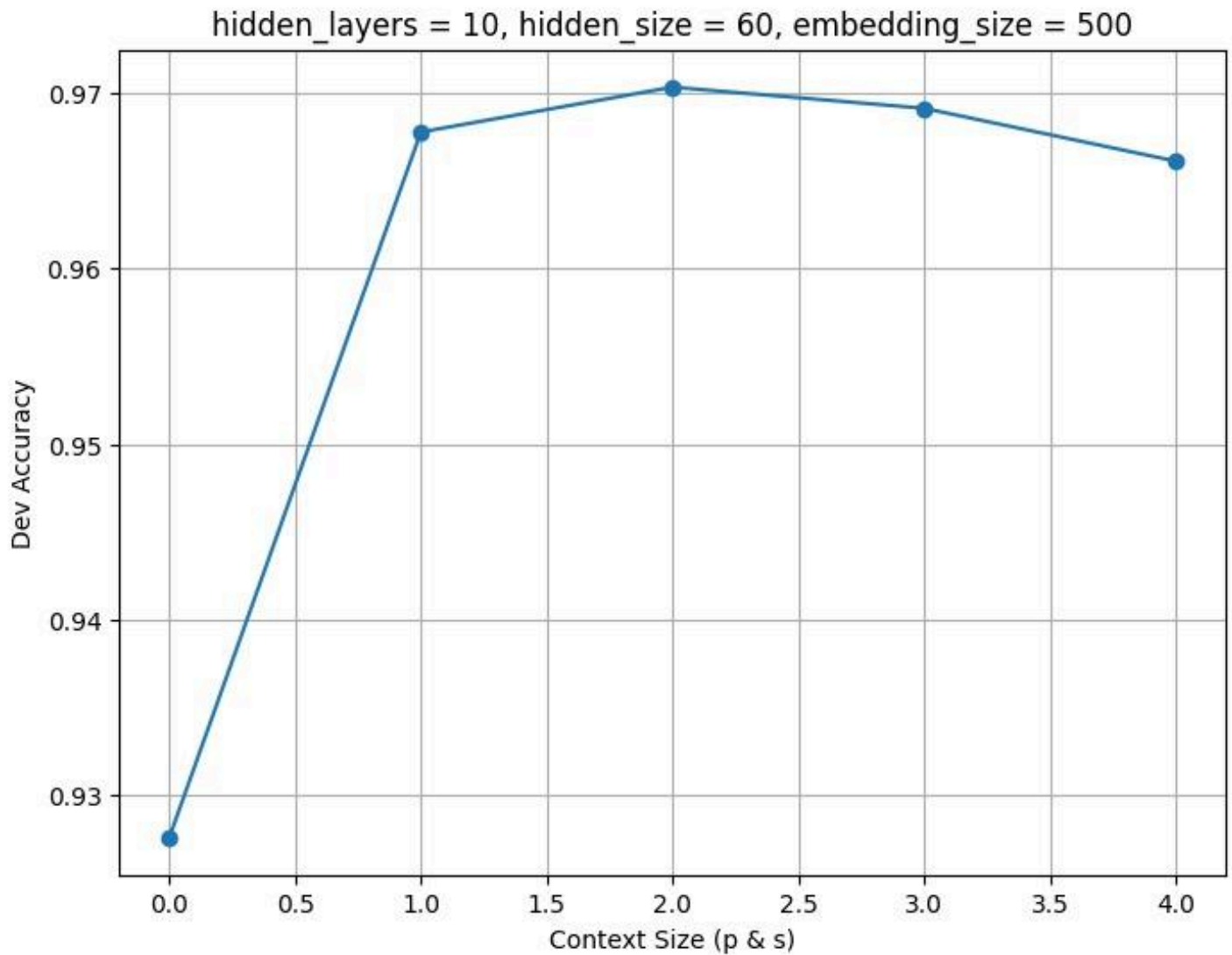
Test Recall: 0.9897

Test F1 Score: 0.9948

=====



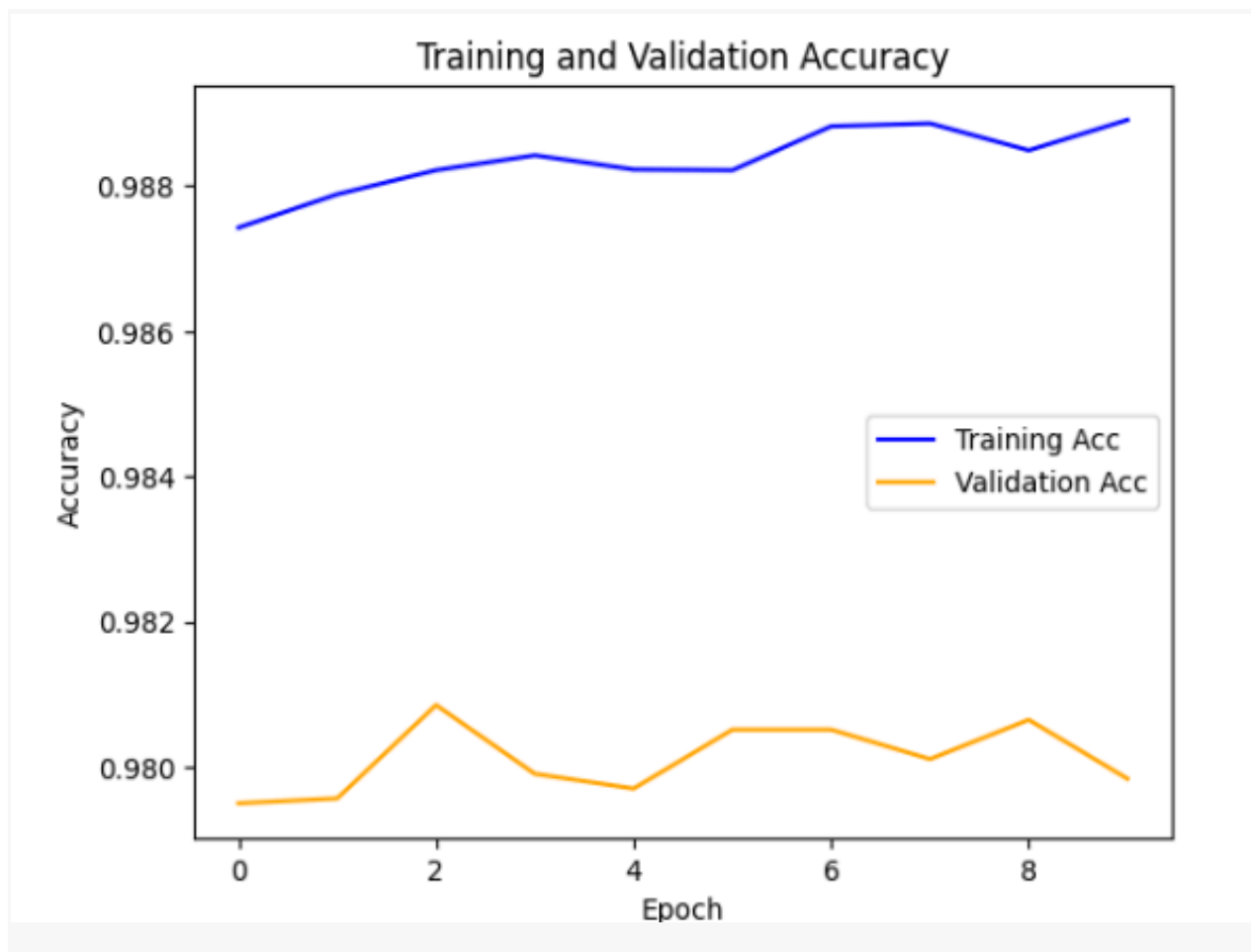
## For Various P and S for configuration:-



## 2. LSTM

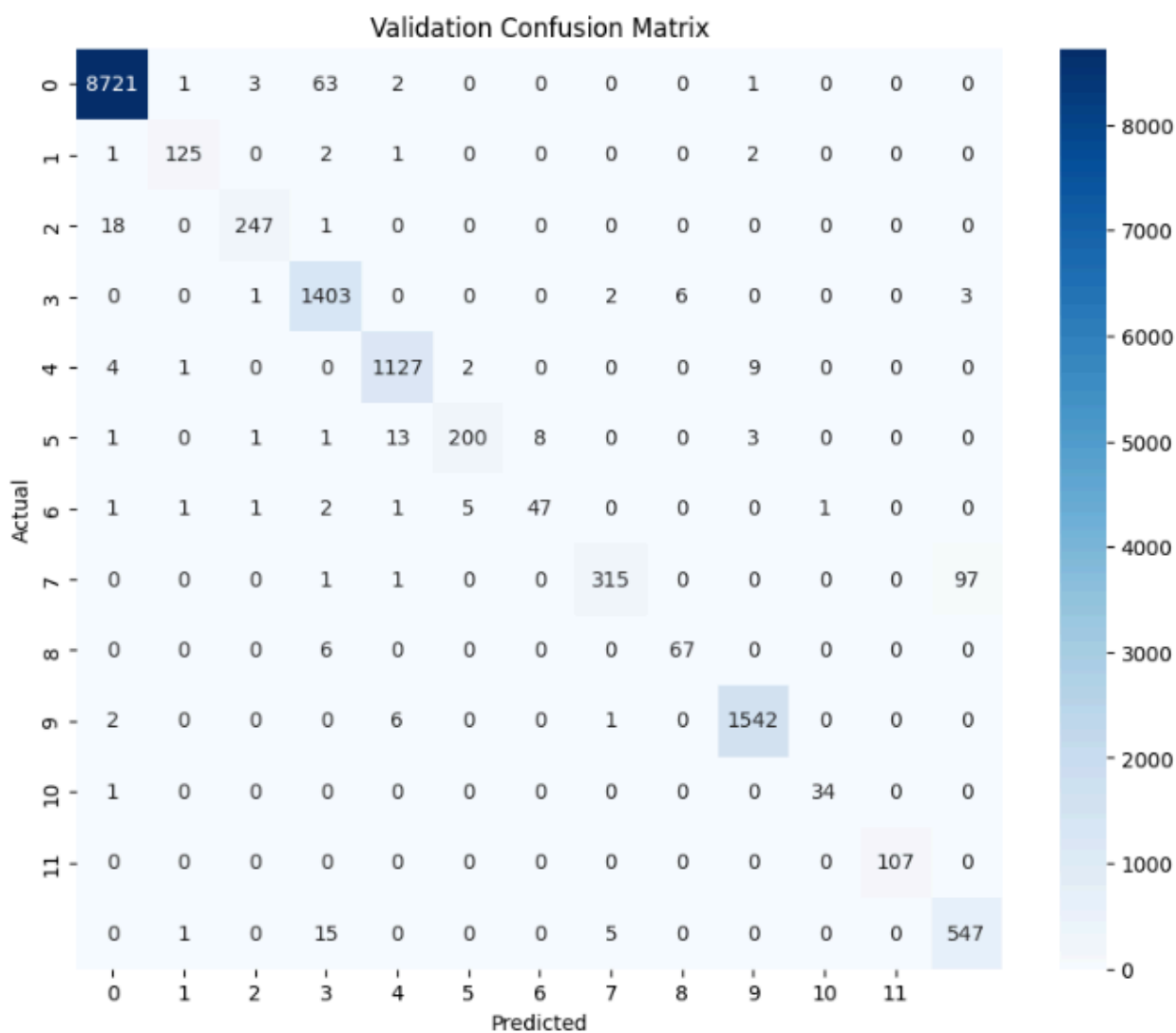
### ● Configuration 1:-

```
BATCH_SIZE = 64  
round_precision = 5  
EPOCHS = 50  
embedding_dim = 100  
hidden_layer_dim = 200  
learning_rate = 1e-3  
epochs = 10  
Bidirectional = False  
Num_layer = 1
```



## For Validation Data :-

```
===== Validation Data =====  
Validation Accuracy: 0.9798  
Validation Precision: 0.9921  
Validation Recall: 0.9921  
Validation F1 Score: 0.9921  
=====
```



## For Test Data:-

===== Test Data =====

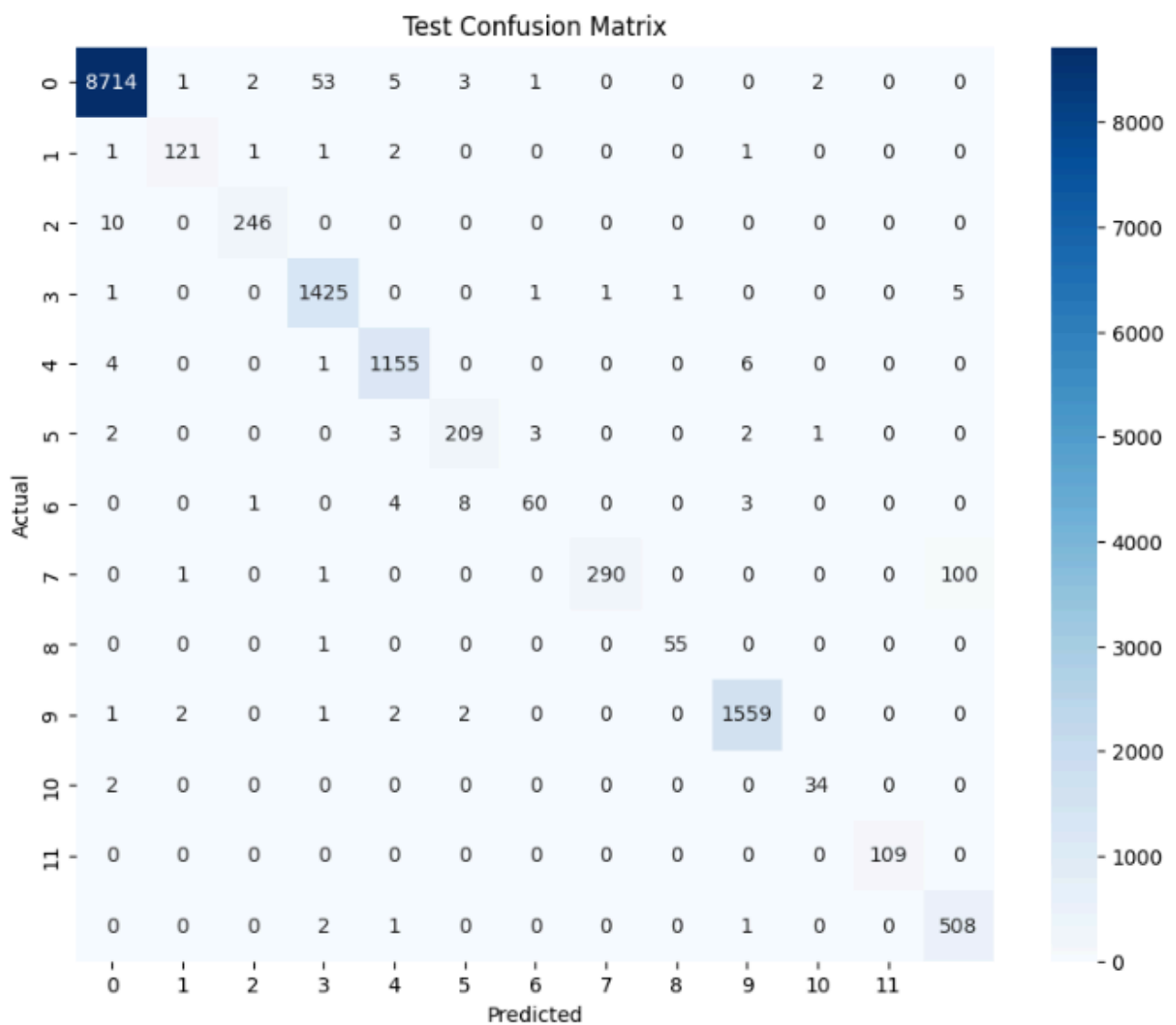
Test Accuracy: 0.9832

Test Precision: 0.9918

Test Recall: 0.9918

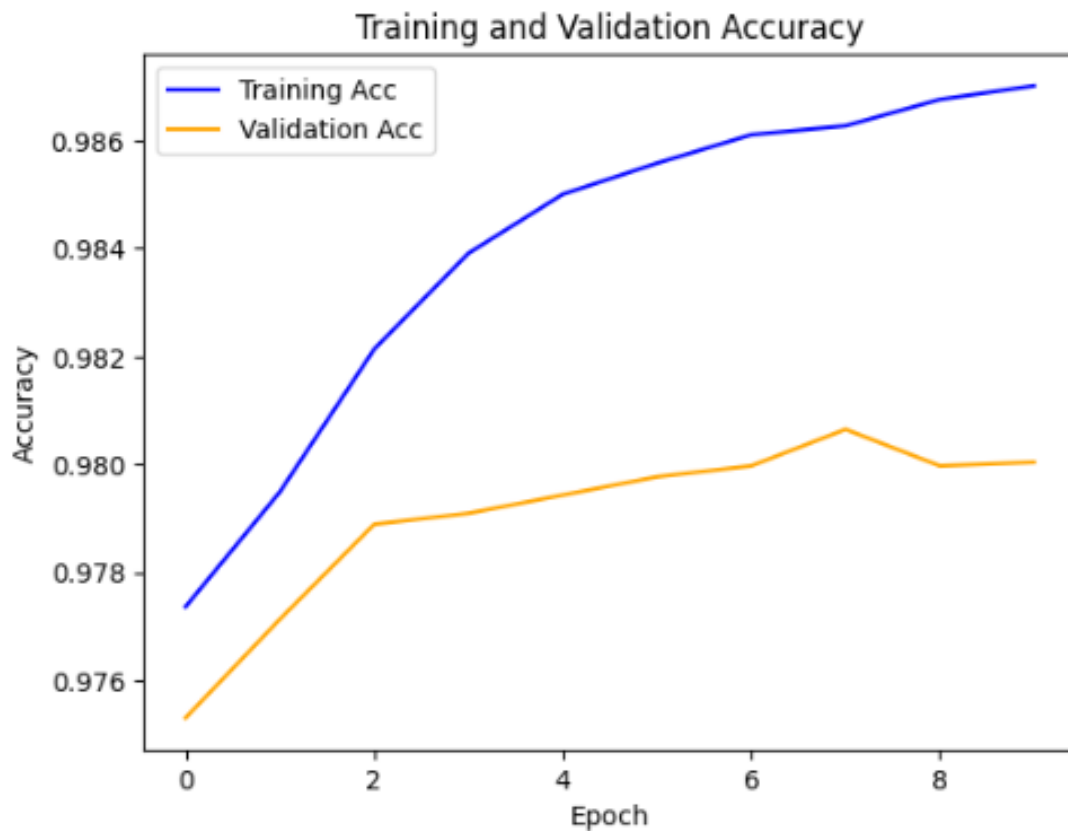
Test F1 Score: 0.9918

=====



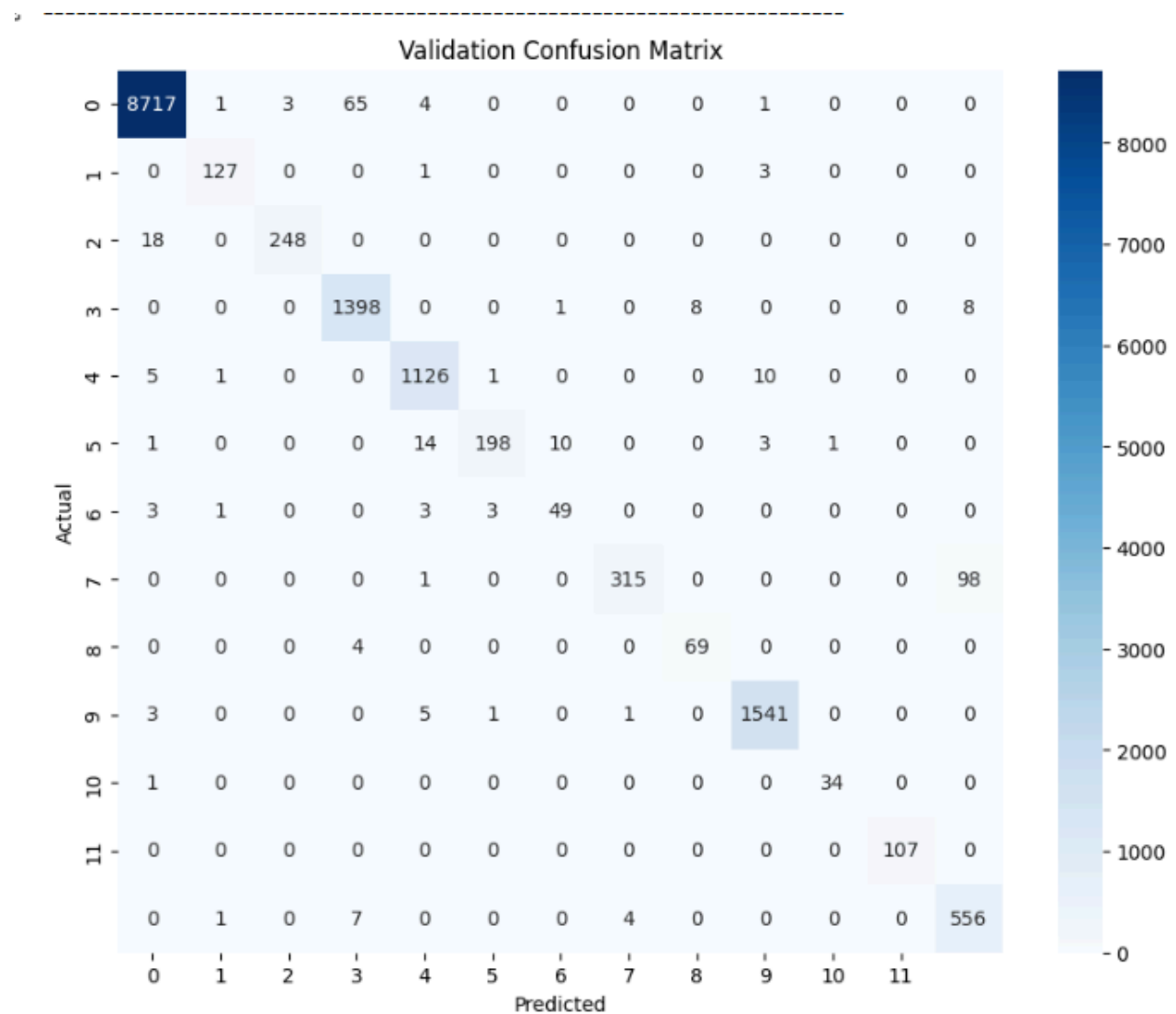
## ● Configuration 2:-

```
BATCH_SIZE = 64
round_precision = 5
EPOCHS = 50
embedding_dim = 100
hidden_layer_dim = 200
learning_rate = 1e-3
epochs = 10
bidirectional = True
num_layer = 2
```



## For Validation Data:-

```
===== Validation Data =====  
Validation Accuracy: 0.9800  
Validation Precision: 0.9922  
Validation Recall: 1.0000  
Validation F1 Score: 0.9961  
=====
```



## For Test Data:-

===== Test Data =====

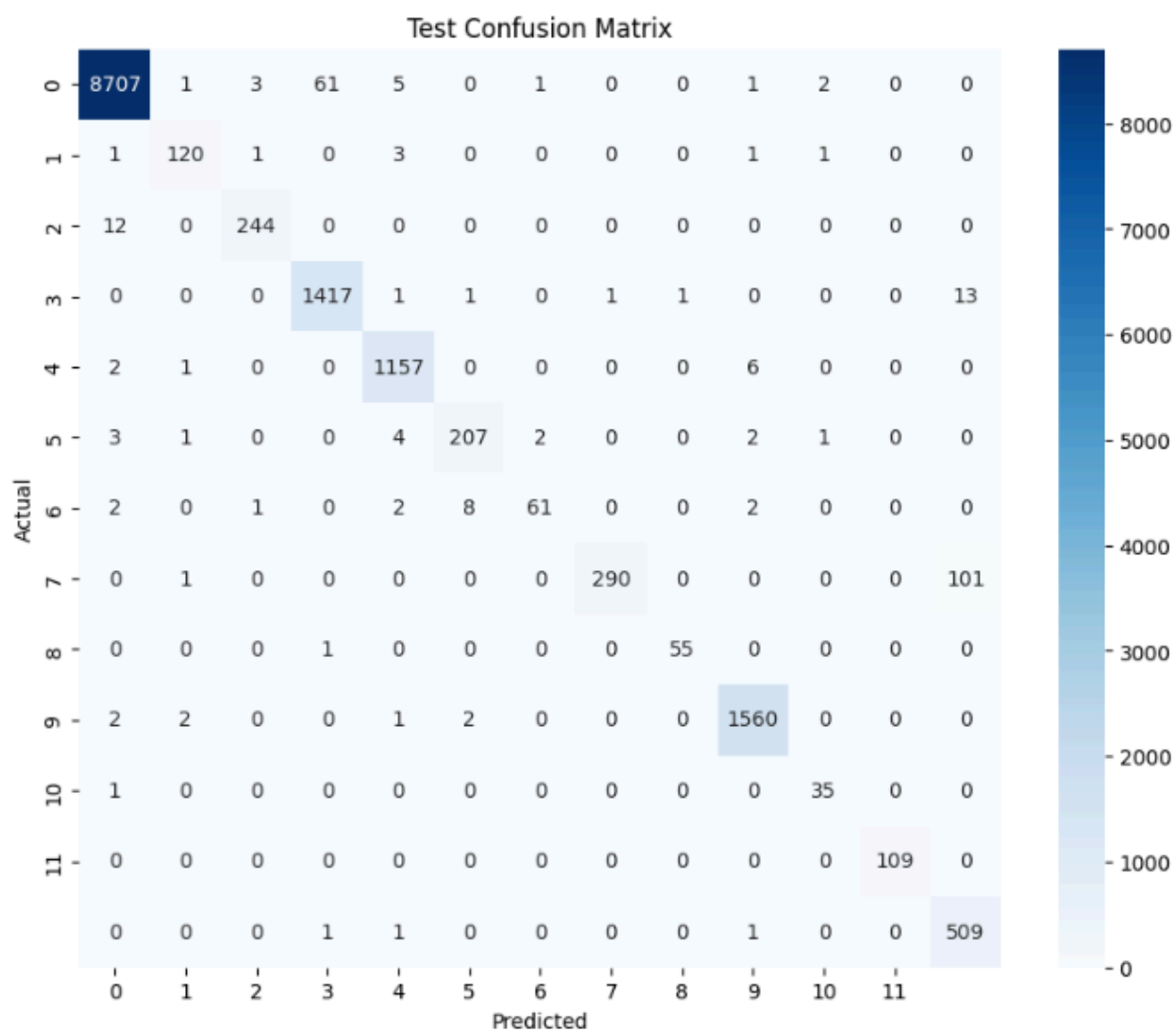
Test Accuracy: 0.9823

Test Precision: 0.9917

Test Recall: 0.9917

Test F1 Score: 0.9917

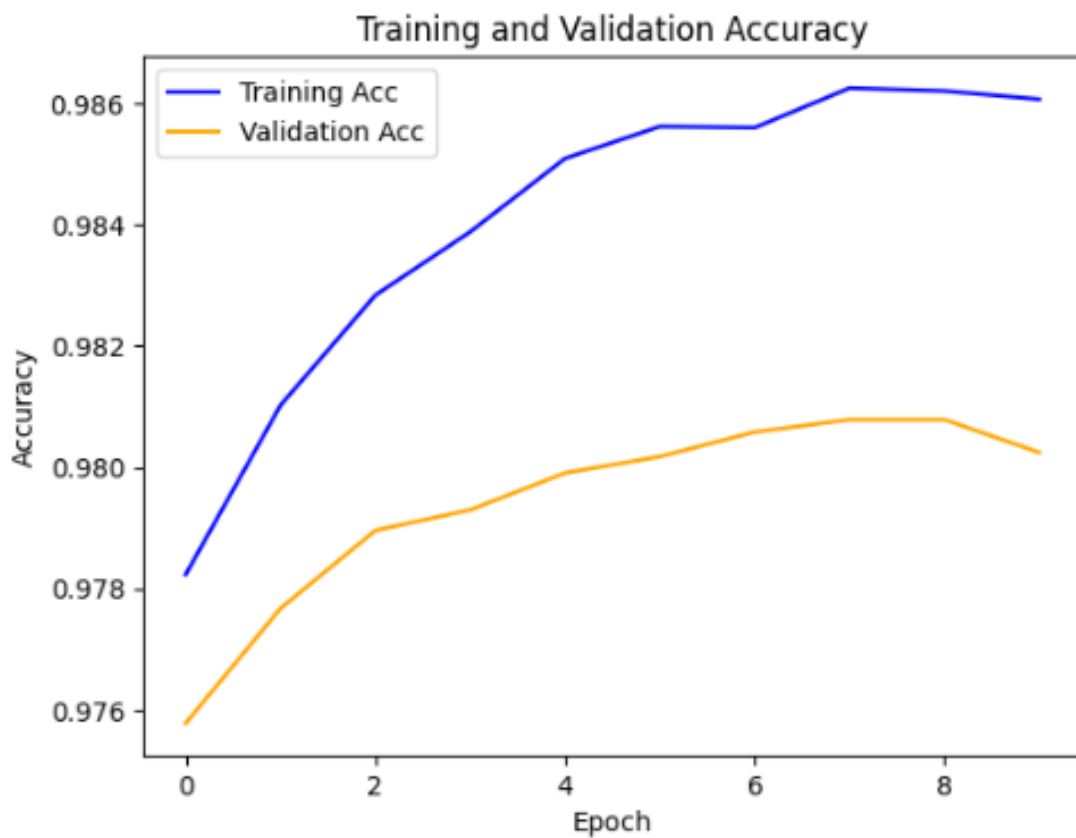
=====



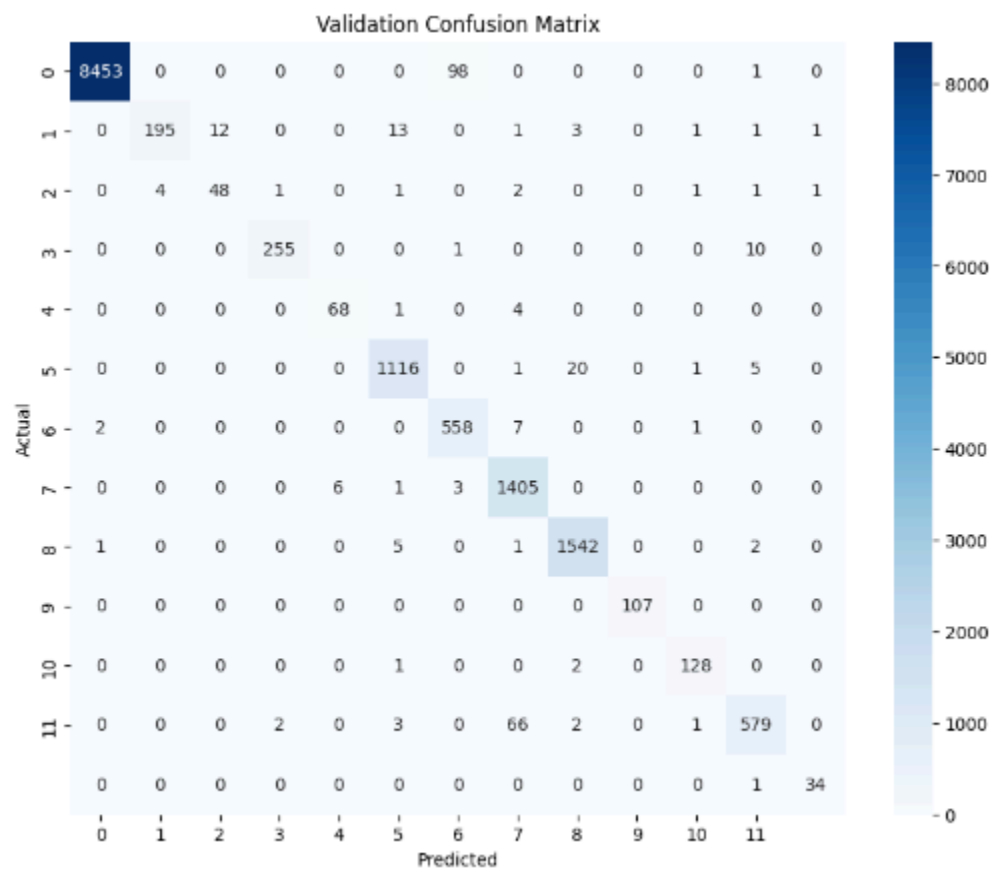


### ● Configuration 3:-

```
BATCH_SIZE = 64  
round_precision = 5  
EPOCHS = 10  
embedding_dim = 300  
hidden_layer_dim = 400  
learning_rate = 1e-3  
epochs = 10  
bidirectional = True  
num_layer = 4
```



```
===== Validation Data =====
Validation Accuracy: 0.9802
Validation Precision: 1.0000
Validation Recall: 1.0000
Validation F1 Score: 1.0000
=====
```



**For Test Data :-**

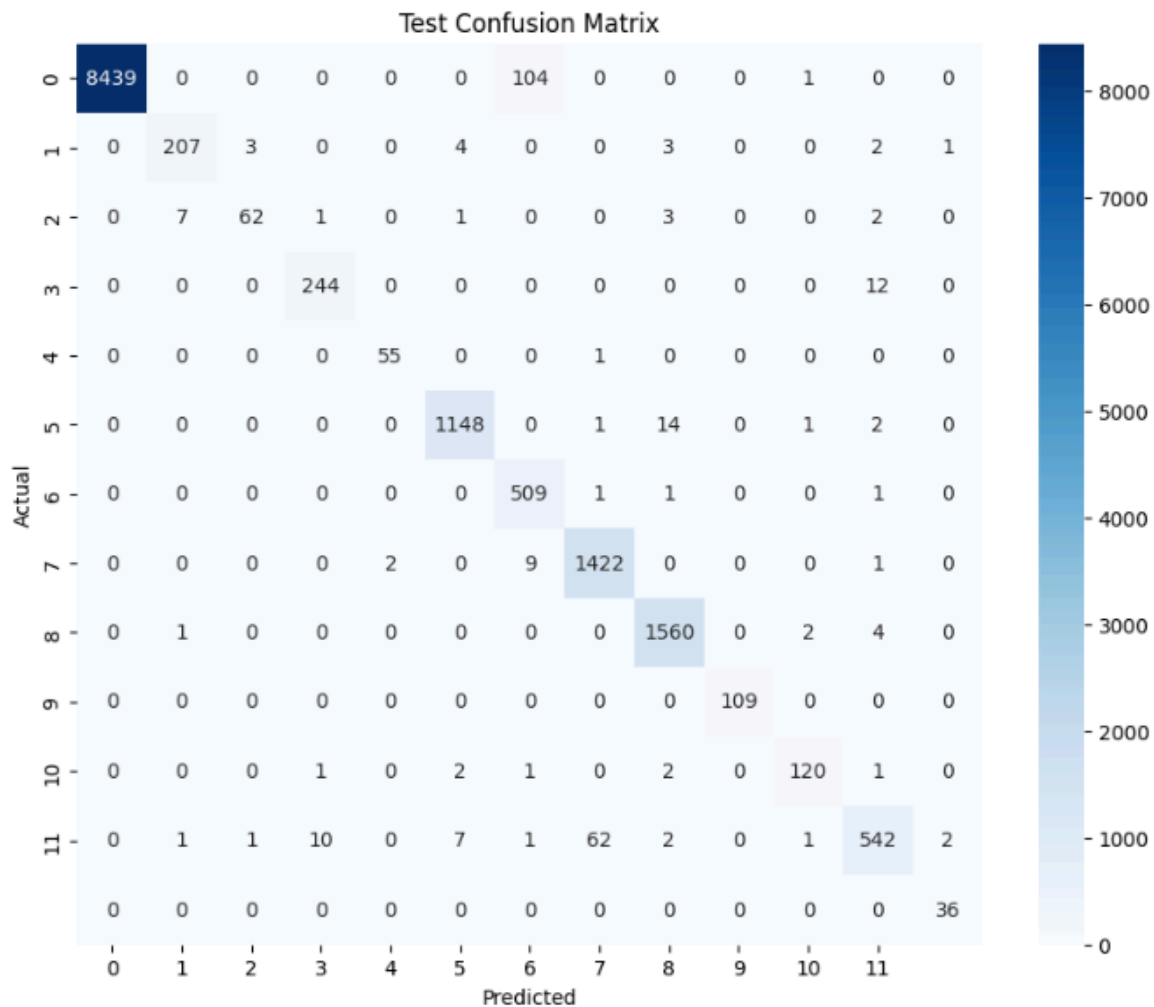
```
===== Test Data =====
```

Test Accuracy: 0.9811

Test Precision: 1.0000

Test Recall: 1.0000

Test F1 Score: 1.0000



## ❖ Analysis:-

### ● **Analytical Result of Feed Forward Neural Network POS Tagger:**

- Advancing Model Complexity: By increasing parameters such as embedding dimension and hidden layer sizes, there's a slight uptick in development accuracy, both in macro and weighted averages.
- Test Accuracy Enhancement: Augmenting model complexity also leads to a noticeable boost in test accuracy.
- Influence of Activation Functions: Switching from ReLU to Tanh activation function causes a slight dip in accuracy, favoring ReLU for better performance.
- Context Window Impact:
  - The context window size significantly affects model accuracy.
  - The lowest accuracy is observed when both preceding and succeeding window sizes are 0, while the highest accuracy is achieved when both are 1.
  - Further increase in window size results in a minor decrease in accuracy.
- Activation function: ReLu gives the best performance and leaky ReLu gives slightly less good performance than ReLu and softmax gives even more noticeable lesser accuracy

### ● **Analytical Result of LSTM (RNN) POS Tagger:**

- Effect of Layer Count: Increasing the number of layers extends training time without notable accuracy improvement.
- Bidirectionality Effect: Activating bidirectionality notably boosts accuracy, showcasing its importance in capturing context effectively.

- Epoch Count Impact: Despite increasing epochs, accuracy doesn't improve significantly due to bidirectionality's substantial performance enhancement.
- Activation Functions: Tanh function is utilized internally within LSTM architecture, while SoftMax is applied at the output layer as the final activation function.