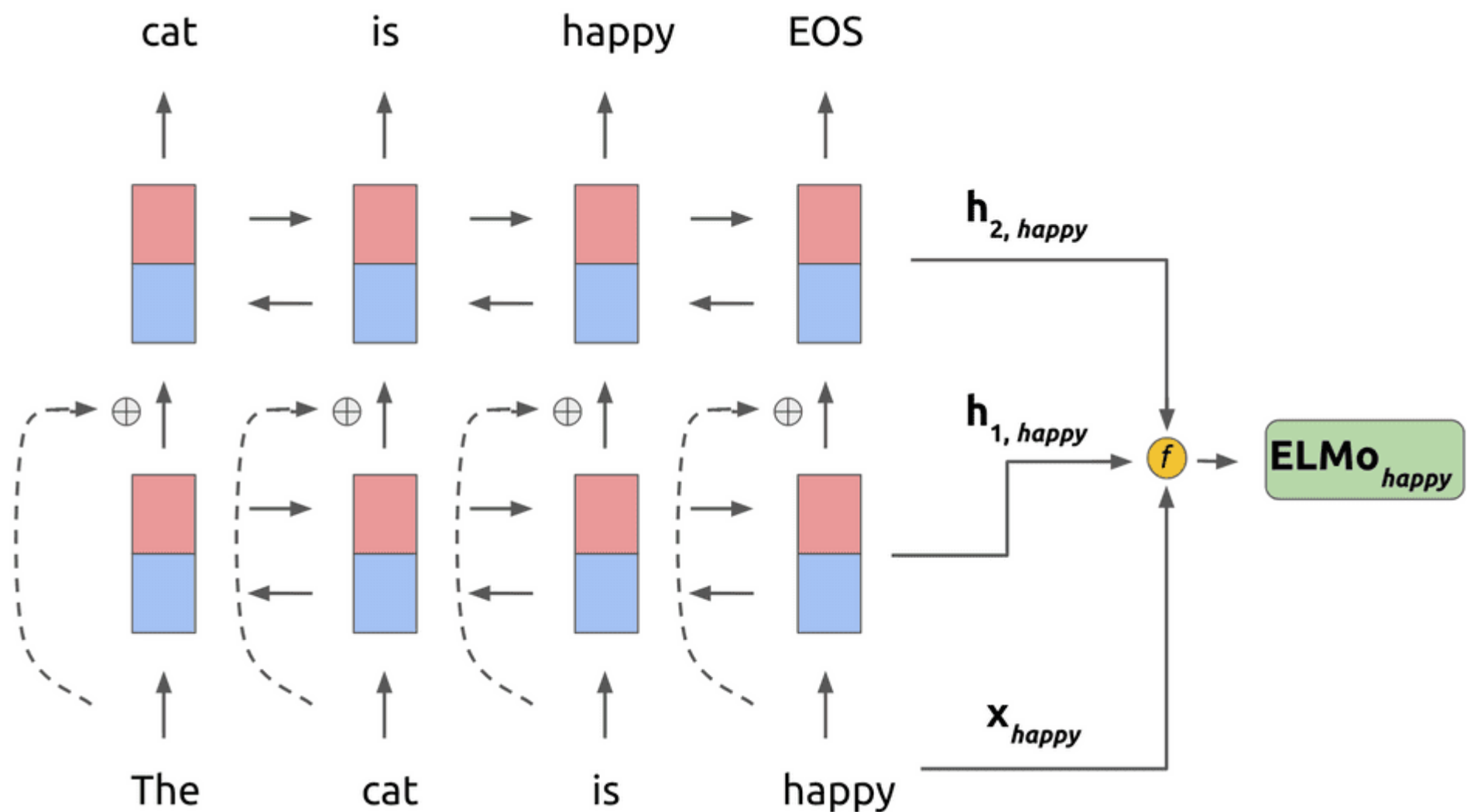


# Assignment 4

## Report

### ELMo Architecture



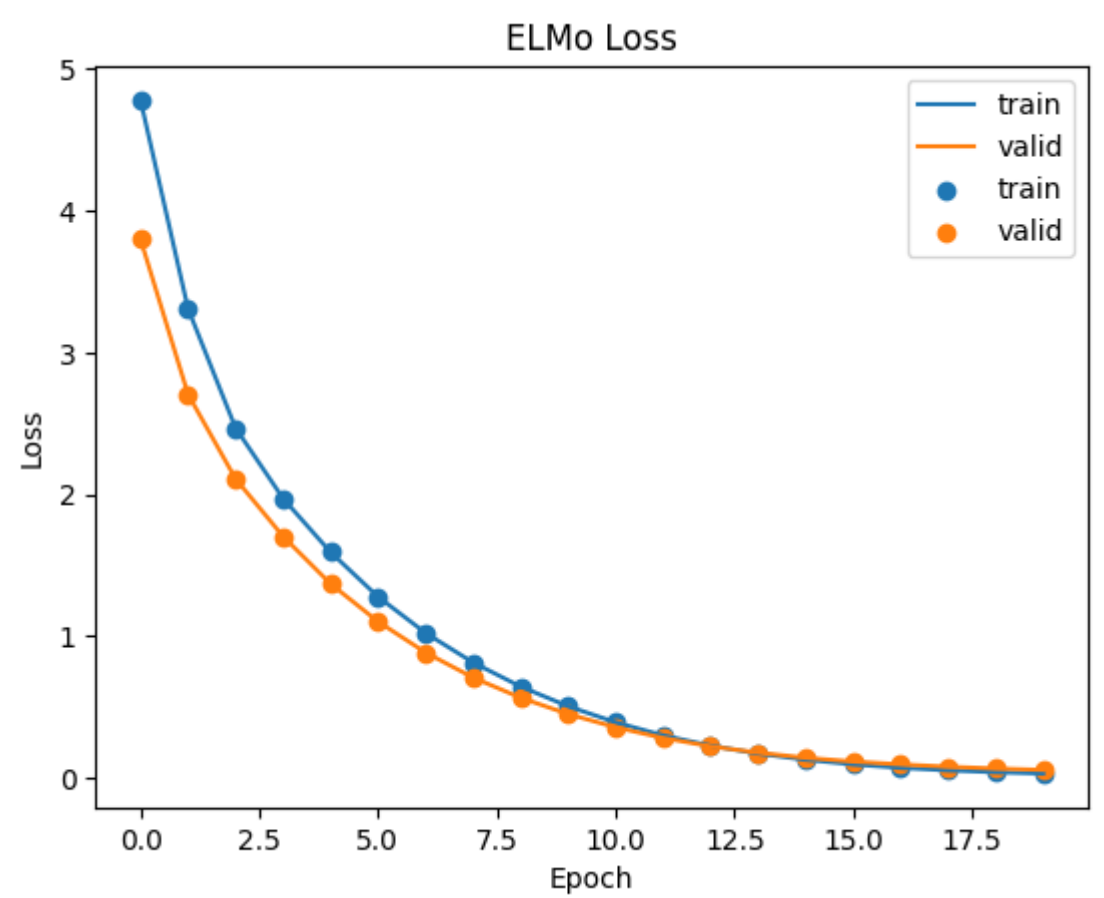
- The model takes in an input tensor of size  $(batch\_size, sequence\_length)$  where each element in the tensor is a word index representing a token in the input text.
- The embedding layer looks up the pretrained GloVe embeddings for each word in the input tensor and returns a tensor of size  $(batch\_size, sequence\_length, embedding\_dim)$  where  $embedding\_dim$  is the size of the GloVe embedding vectors.
- The `lstm1` layer is a bidirectional LSTM layer with  $hidden\_dim$  hidden units, which means it has  $hidden\_dim$  hidden units in both the forward and backward directions. This layer takes in the output of the embedding layer and returns a tensor of size  $(batch\_size, sequence\_length, hidden\_dim*2)$  because the outputs of the forward and backward LSTMs are concatenated along the last dimension.
- The `lstm2` layer is also a bidirectional LSTM layer with  $hidden\_dim$  hidden units that takes in the output of `lstm1` and returns a tensor of size  $(batch\_size, sequence\_length, hidden\_dim*2)$ .
- The `linear_out` layer is a linear layer that takes in the output of `lstm2` and returns a tensor of size  $(batch\_size, sequence\_length, vocab\_size)$  where  $vocab\_size$  is the number of output classes (i.e. the size of the output vocabulary).

Overall, the model is designed to take in a sequence of word indices, embed those words using pretrained GloVe embeddings, and then use a bidirectional LSTM to encode the sequence into a fixed-length vector representation that can be used for classification or other downstream tasks.

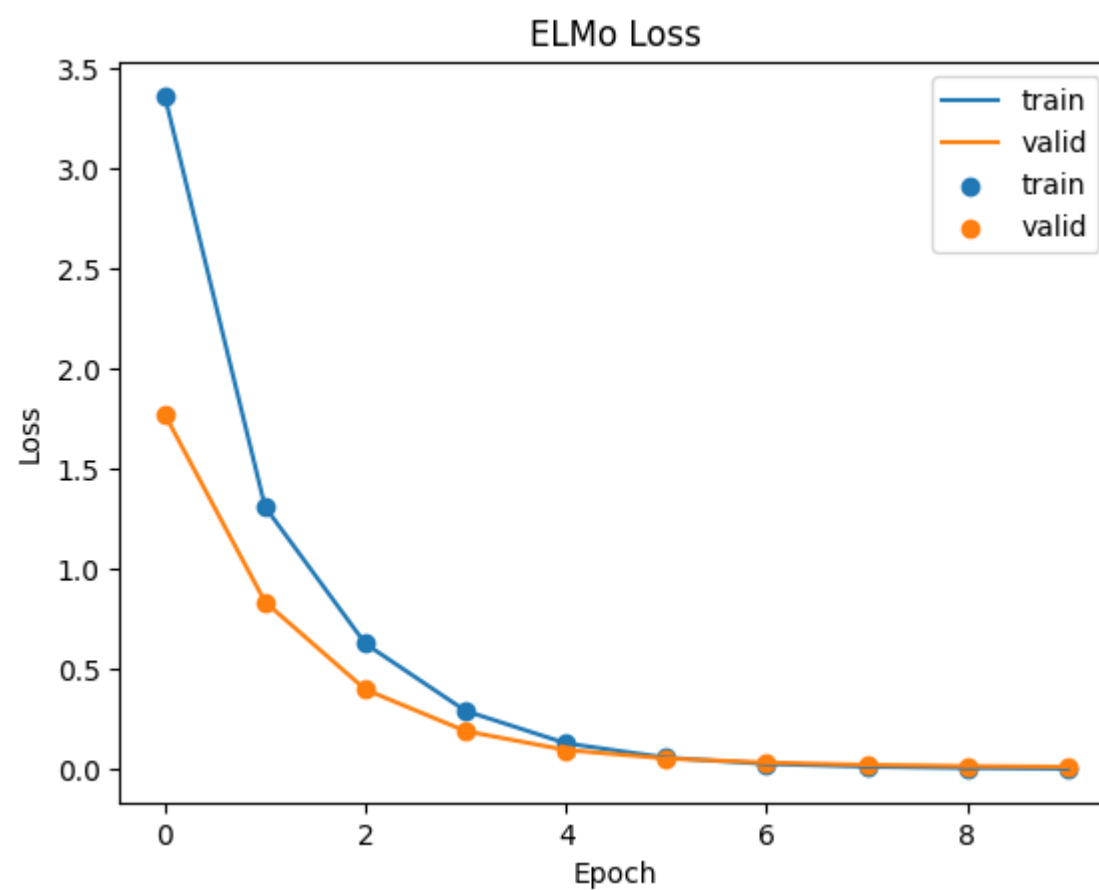
```
ELMo(  
  (embedding): Embedding(10394, 300, padding_idx=0)  
  (lstm1): LSTM(300, 100, batch_first=True, bidirectional=True)  
  (lstm2): LSTM(200, 100, batch_first=True, bidirectional=True)  
  (linear_out): Linear(in_features=200, out_features=10394, bias=True)  
)
```

The inputs are the the sequence of word indices from 0 to length-1 , and the output is the sequence of word indices from 1 to length .

**Dataset : Stanford Sentiment Treebank**



**Dataset : Multi-Genre NLP corpus**



## Classification

The architecture consists of the following layers:

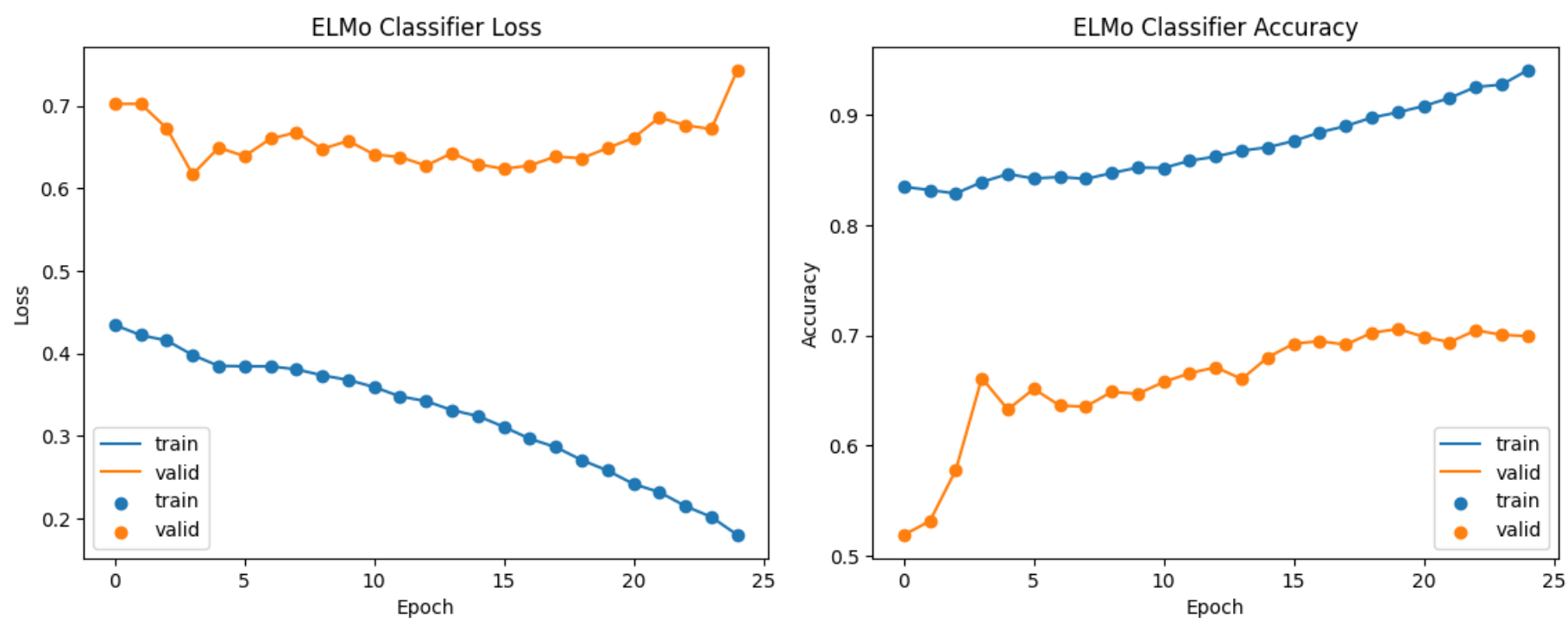
- `nn.Embedding`: The input is first passed through an embedding layer using the pretrained ELMo embedding.  
`nn.Linear`: The output from the embedding layer is passed through a linear layer to change the embedding dimension to hidden dimensionx2.
- `nn.LSTM`: The resulting tensor is then passed through two layers of LSTM in a sequential manner.  
`torch.max`: After passing through the LSTM layers, the resulting tensor is passed through a max pooling layer.
- `nn.Dropout`: The max pooling output is then passed through a dropout layer with a dropout probability of 0.5.

`nn.Linear`: Finally, the resulting tensor is passed through a linear layer to get the output of the classifier.

This architecture also includes `L2 regularization`.

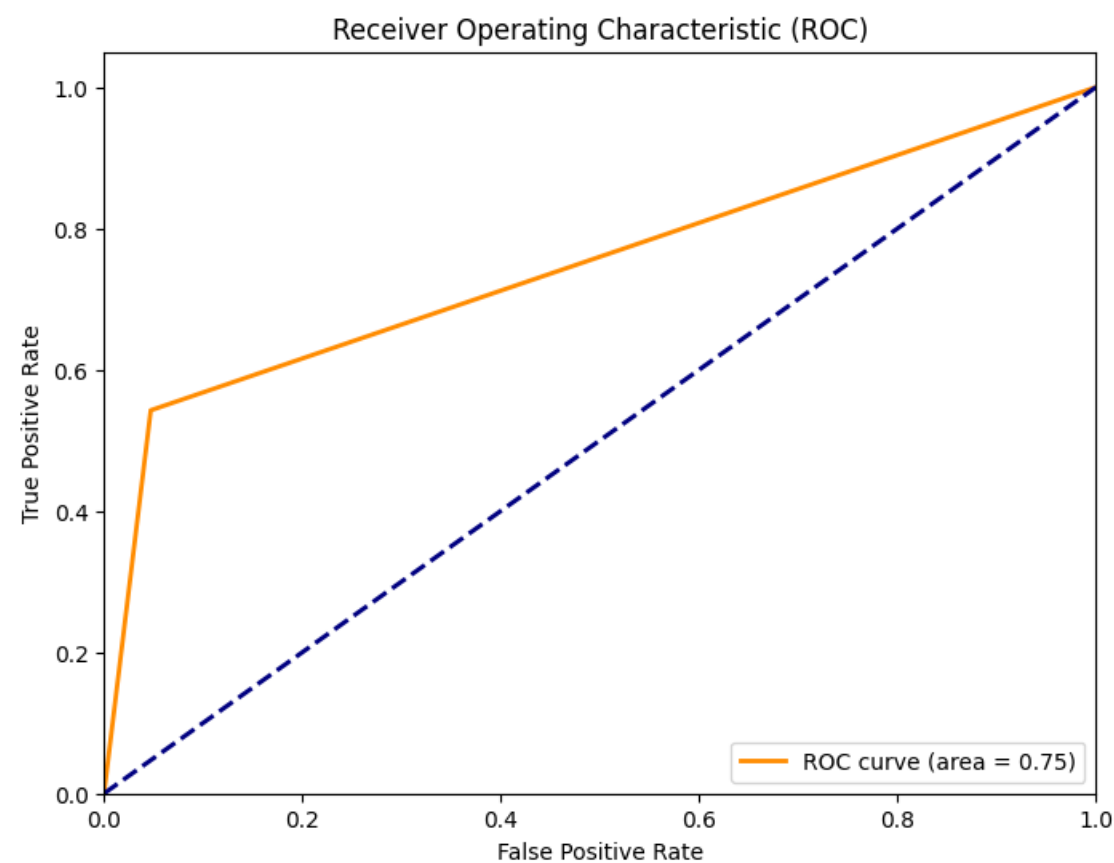
```
scoreClassifier(
(embeddings): Embedding(10394, 300, padding_idx=0)
(lstm1): LSTM(300, 100, batch_first=True, bidirectional=True)
(lstm2): LSTM(200, 100, batch_first=True, bidirectional=True)
(linear1): Linear(in_features=300, out_features=200, bias=True)
(linear_out): Linear(in_features=200, out_features=3, bias=True)
(dropout): Dropout(p=0.5, inplace=False)
)
```

**Dataset : Stanford Sentiment Treebank**



Visualizations

Training Dataset

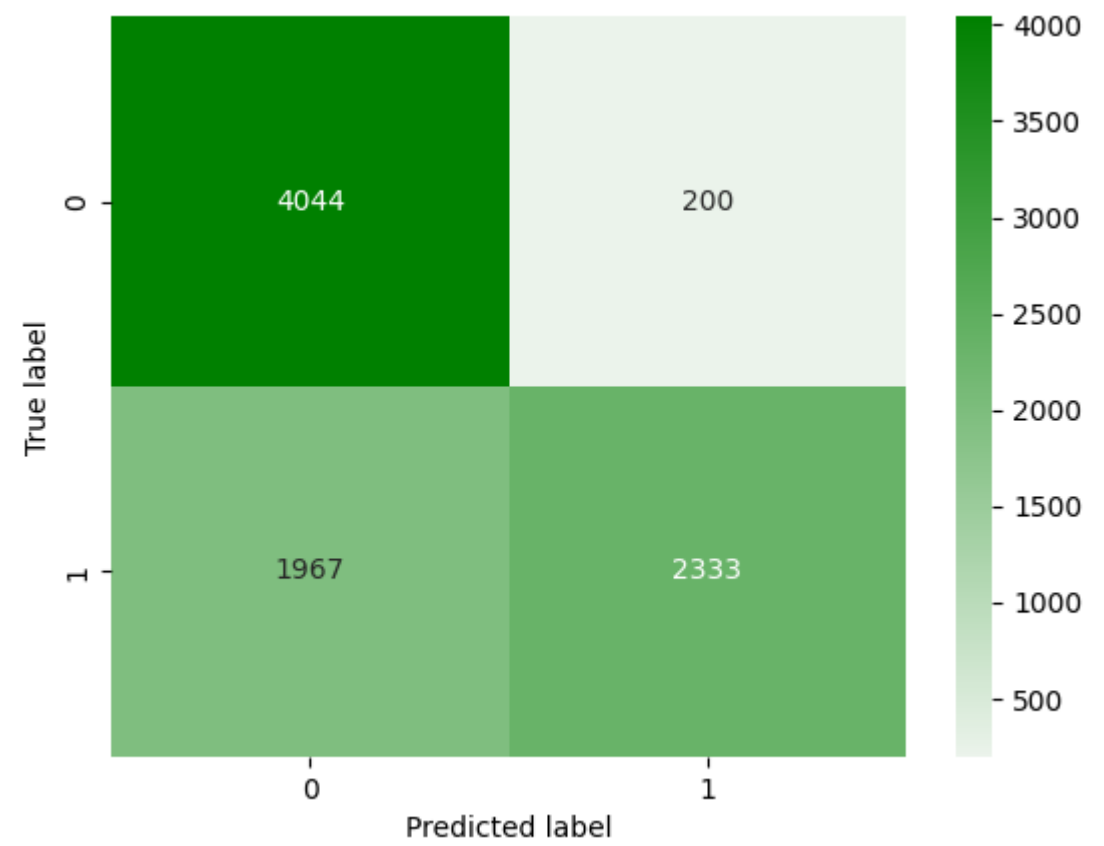


Training Dataset

	precision	recall	f1-score	support
0	0.67	0.95	0.79	4244
1	0.92	0.54	0.68	4300
accuracy			0.75	8544
macro avg	0.80	0.75	0.74	8544
weighted avg	0.80	0.75	0.74	8544

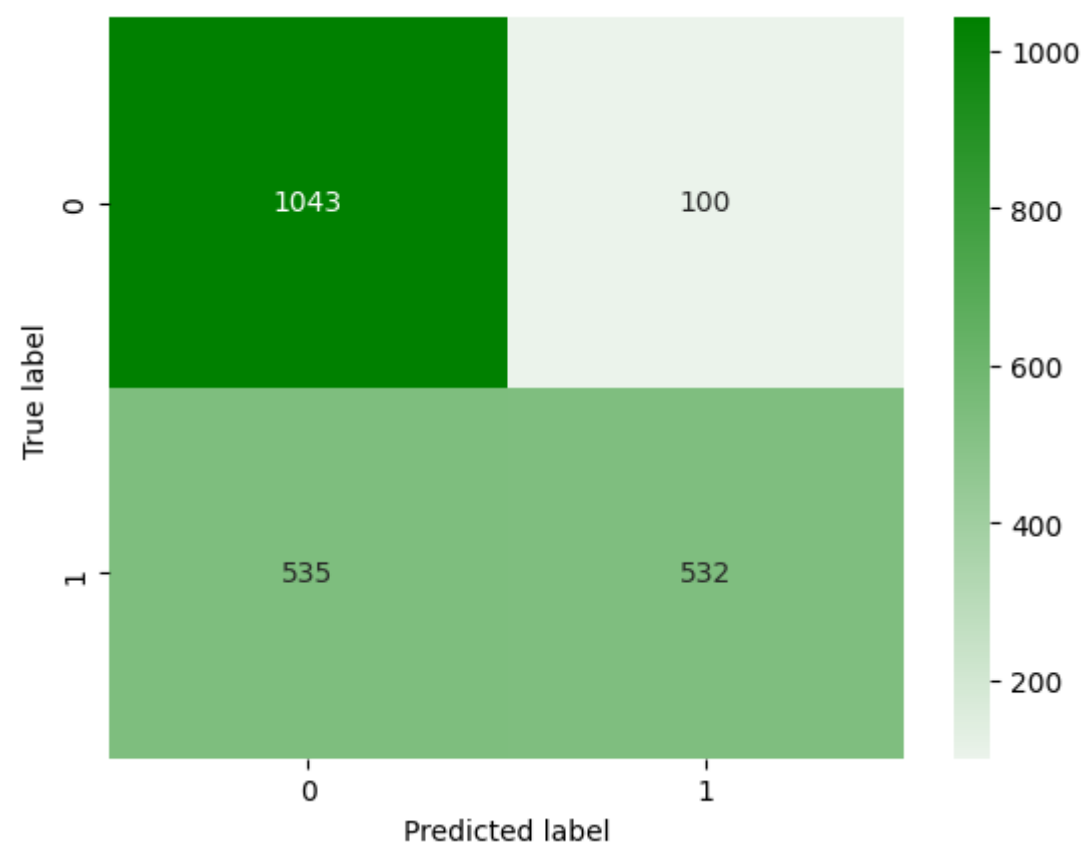
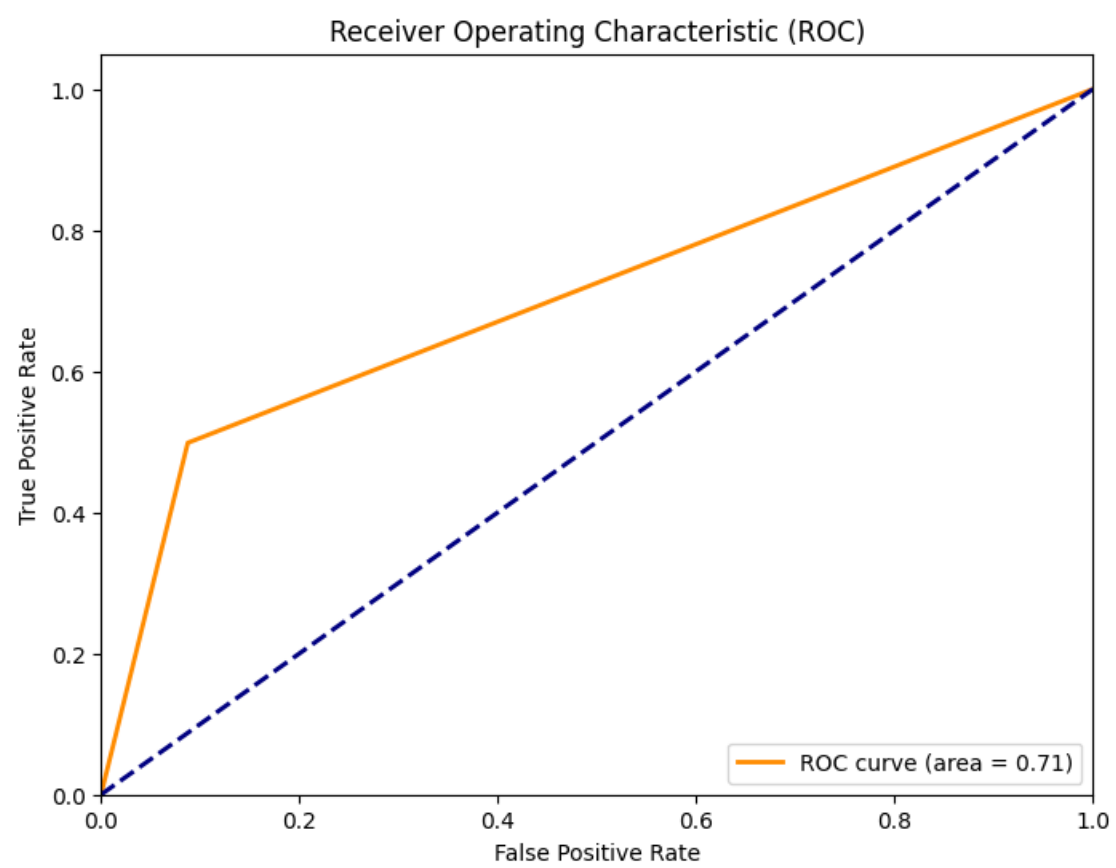
Accuracy: 0.7463717228464419  
F1 Score: 0.6828625786623739  
Precision: 0.9210422424003158

Recall: 0.5425581395348837  
Confusion Matrix:  
[[4044. 200.]  
 [1967. 2333.]]

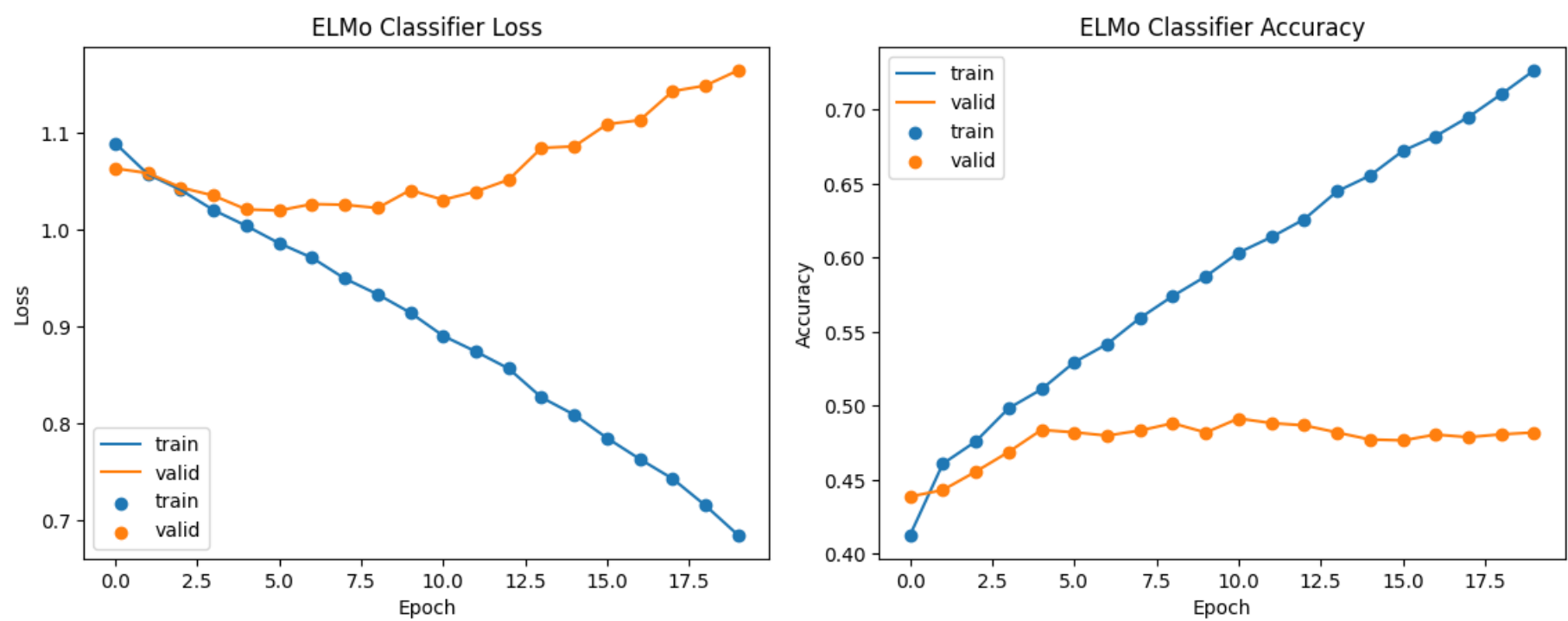


Testing Dataset

	precision	recall	f1-score	support
0	0.66	0.91	0.77	1143
1	0.84	0.50	0.63	1067
accuracy			0.71	2210
macro avg	0.75	0.71	0.70	2210
weighted avg	0.75	0.71	0.70	2210
Accuracy: 0.7126696832579186				
F1 Score: 0.6262507357268982				
Precision: 0.8417721518987342				
Recall: 0.4985941893158388				
Confusion Matrix:				
[[1043. 100.]				
[ 535. 532.]]				



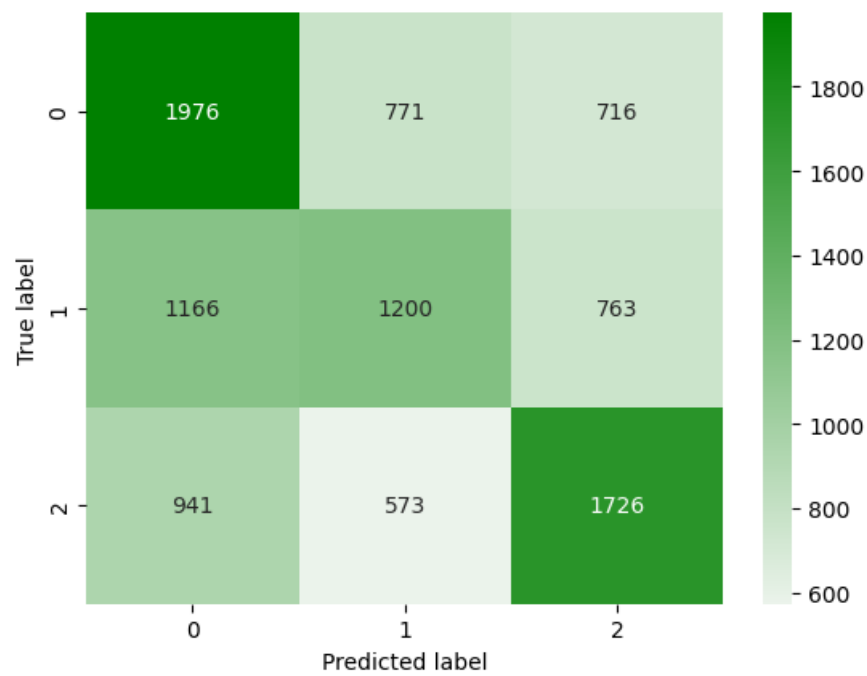
Dataset : Multi-Genre NLP corpus



Results:

Testing

	precision	recall	f1-score	support
negative	0.48	0.57	0.52	3463
neutral	0.47	0.38	0.42	3129
positive	0.54	0.53	0.54	3240
accuracy			0.50	9832
macro avg	0.50	0.50	0.49	9832
weighted avg	0.50	0.50	0.50	9832
Accuracy: 0.49857607811228644				
Confusion Matrix:				
[[1976. 771. 716.]				
[1166. 1200. 763.]				
[ 941. 573. 1726.]]				



Training

	precision	recall	f1-score	support
negative	0.78	0.87	0.82	8565
neutral	0.85	0.77	0.81	7525
positive	0.85	0.83	0.84	9134
accuracy			0.82	25224
macro avg	0.83	0.82	0.82	25224
weighted avg	0.83	0.82	0.82	25224

Accuracy: 0.8242150333016175

Confusion Matrix:

```
[[7418.  550.  597.]
 [ 990. 5784.  751.]
 [1057.  489. 7588.]]
```

