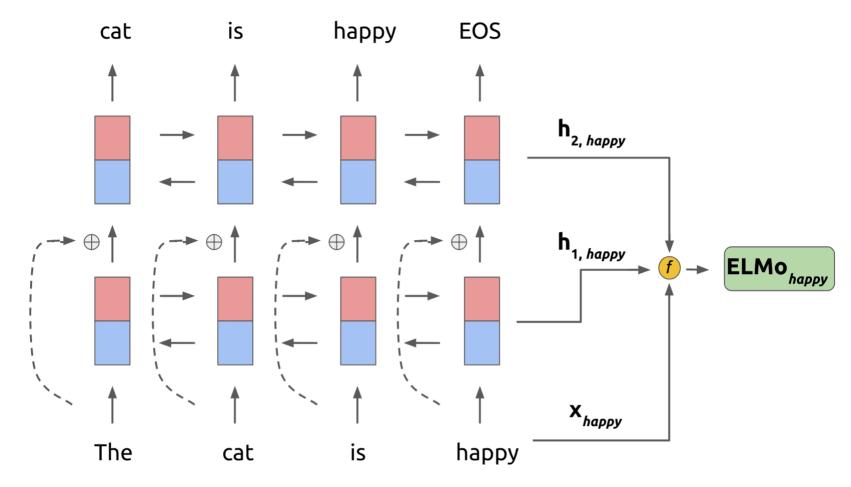
Assignment 4

Report

ELMo Architecture



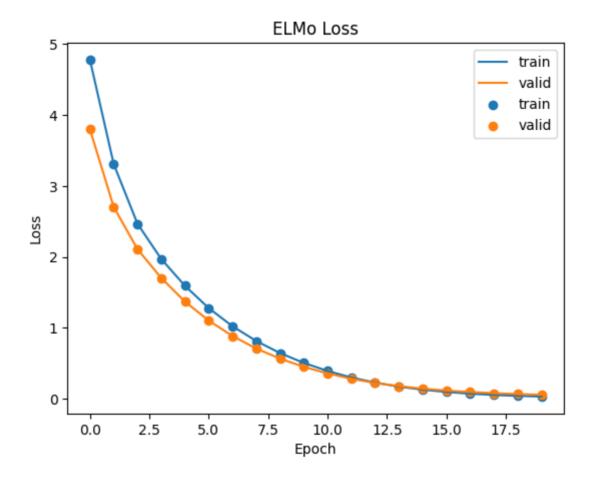
- The model takes in an input tensor of size (batch_size, sequence_length) where each element in the tensor is a word index representing a token in the input text.
- The embedding layer looks up the pretrained GloVe embeddings for each word in the input tensor and returns a tensor of size (batch_size, 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.

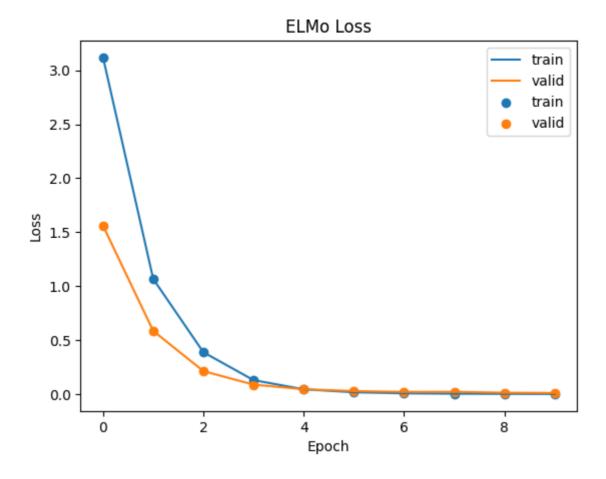
```
ELMo(
  (embedding): Embedding(10942, 300, padding_idx=0) // Glove pretrained embedding
  (lstm1): LSTM(300, 100, batch_first=True, bidirectional=True)
  (lstm2): LSTM(200, 100, batch_first=True, bidirectional=True)
  (linear1): Linear(in_features=300, out_features=100, bias=True)
  (linear_out): Linear(in_features=200, out_features=10942, 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

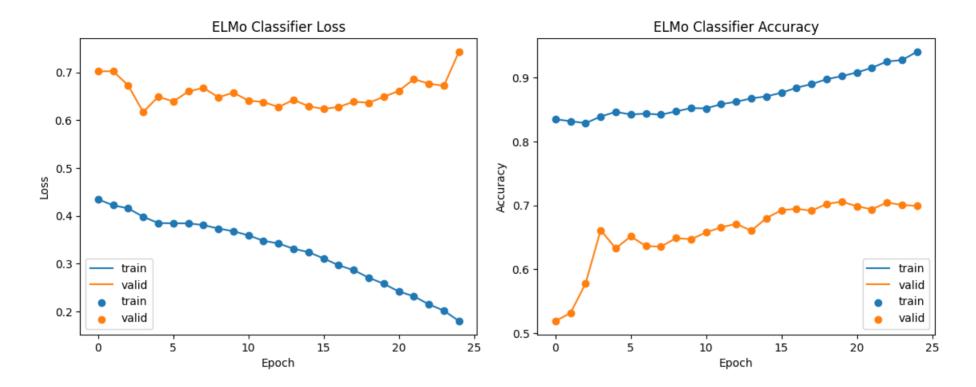
The architecture consists of the following layers:

- 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 dimension*2.
- 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.

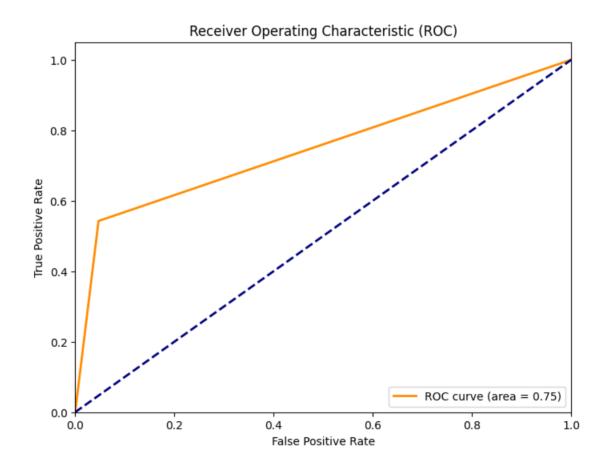
This architecture also includes L2 regularization.

```
scoreClassifier(
  (embeddings): Embedding(10942, 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

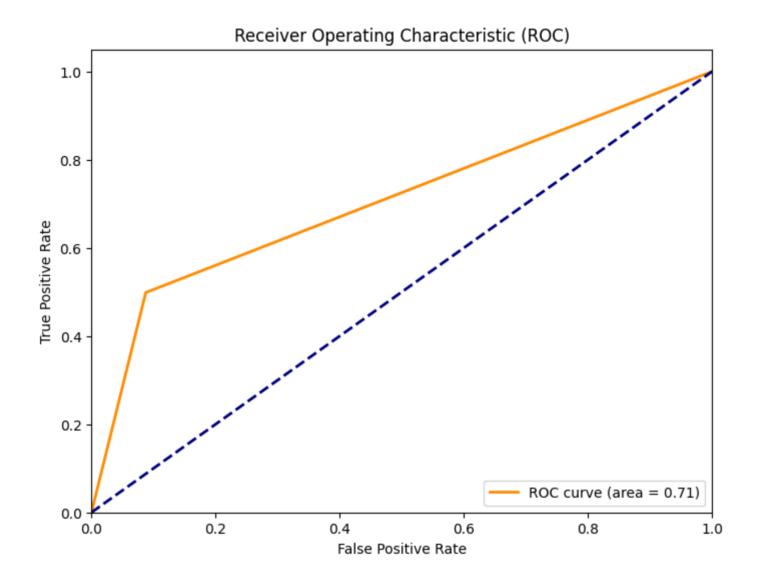


Training Dataset		precision	recall	f1-score	support	
	Θ	0.67	0.95	0.79	4244	
	1	0.92	0.54	0.68	4300	
accur	асу			0.75	8544	
macro	avg	0.80	0.75	0.74	8544	
weighted	avg	0.80	0.75	0.74	8544	

Accuracy: 0.7463717228464419 F1 Score: 0.6828625786623739 Precision: 0.9210422424003158 Recall: 0.5425581395348837

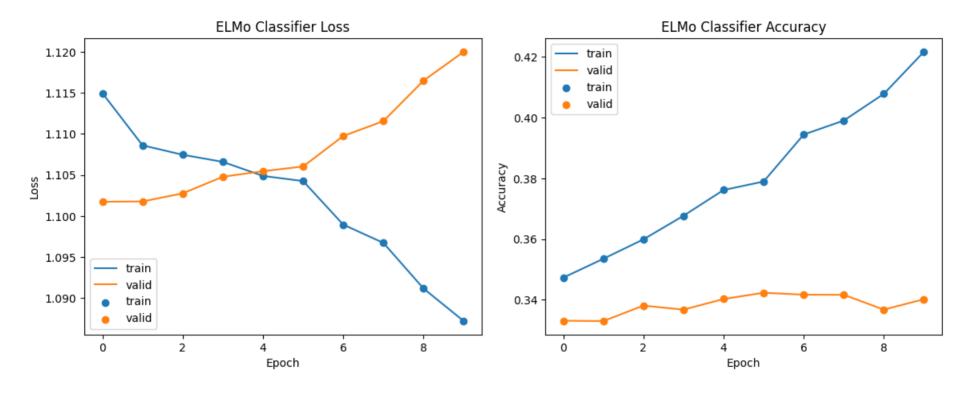
Confusion Matrix: [[4044. 200.] [1967. 2333.]]

Test Dataset



	precision	recall	f1-score	support
Θ	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
	F1 Score: Precision	0.626250 n: 0.8417 0.49859418 n Matrix:	96832579186 97357268982 72151898734 893158388	

Dataset: Multi-Genre NLP corpus



Visualizations

Training Dataset								
	precision	recall	f1-score	support				
	negative	0.37	0.21	0.27	10371			
	neutral	1.00	0.00	0.00	9130			
	positive	0.37	0.82	0.51	11105			
	·							
	accuracy			0.37	30606			
	macro avg	0.58	0.34	0.26	30606			
	weighted avg	0.56	0.37	0.27	30606			
Accuracy: 0.3679997386133438								
Confusion Matrix:								
Confusion Matrix:								
	[[2:	153. 0.	8218.]					
	[16	645. 0.	7485.]					
	Γ19	995. 0.	9110.]]					
	[01	00.]]					

Testing Dataset						
р	recision reca	ll f1-sco	ore suppo	ort		
	negative	0.35	0.14	0.20	3463	
	neutral	1.00	0.00	0.00	3129	
	positive	0.33	0.86	0.48	3240	
	accuracy			0.33	9832	

macro	avg	0.56	0.33	0.23	9832
weighted	avg	0.55	0.33	0.23	9832

Accuracy: 0.33329943043124494

Confusion Matrix:

[[492. 0. 2971.] [466. 0. 2663.] [455. 0. 2785.]]