NLP Assignment 2 Report

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Task 1: NER Dataset

The final processed data looks like for the NER Dataset.

This is a screenshot of one of the testing data.

```
'b0311cba3aac4d909eec6e156c059617": {
    "text": "(See Principles of Statutory Interpretation by Justice G.P. Singh, 9th Edn., 2004 at p. \n\n 438.).",
       "O",
       "B_JUDGE",
       "I_JUDGE",
       "0",
"0",
"0",
```

This is a screenshot of an instance of the training data.

```
5c57ba5b84c6fa46bee65e6616633": {

"text": "He was also asked whether Agya <span class=\"hidden_text\" id=\"span_5\"> CRA No.326-D8 of 1998 6</span> Kaur, mother-in-law of the deceased lived separately from Tarlochan Singh.",

"O",

"O",

"O",
      "B_OTHER_PERSON",
"I_OTHER_PERSON"
```

This is a screenshot of an instance of the validation.json file.

Graphs:

Points to note: Training and Validation losses usually decrease until an epoch, and then, training loss drops and tends to become constant while validation losses keep growing. This is basically the point where overfitting happens.

Underfitting: When the training dataset is too complex for our model. It can't capture them. The gap between the curve basically means that, the model is missing something

Task 2: ATE Dataset

```
"1": {
    "text": "Boot time is super fast , around anywhere from 35 seconds to 1 minute .",
    "labels": [
        "B",
        "I",
        "0",
        "0",
        "0",
        "0",
        "0",
        "0",
        "0",
        "0",
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```

This is a screenshot of one of the instances of the testing data.

```
"1": {
    "text": "I charge it at night and skip taking the cord with me because of the good battery life .",
    "labels": [
        "0",
        "0",
        "0",
        "0",
        "0",
        "0",
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        "0",
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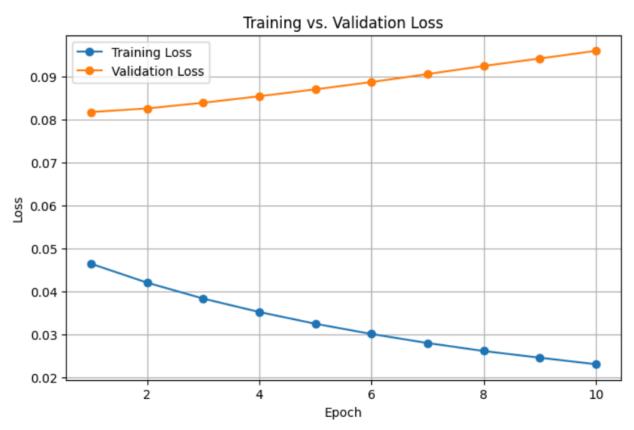
This is a screenshot of an instance of the training data.

```
"2": {
    "text": "This laptop meets every expectation and Windows 7 is great !",
    "labels": [
        "0",
        "0",
        "0",
        "0",
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```

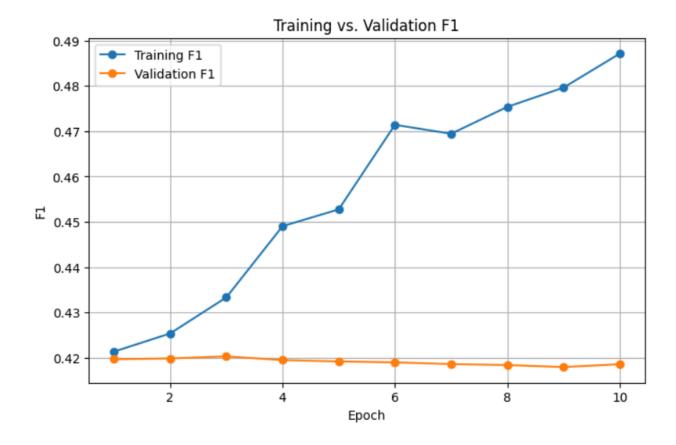
This is a screenshot of an instance of the validation data.

Word2Vec

Vanilla Rnn:



The training loss (blue line) steadily decreases as the number of epochs increases, indicating the model is effectively learning from the training data. However, the validation loss (orange line) remains flat, suggesting the model might be overfitting to the training data. While the model exhibits a decreasing trend in training loss, there is a risk of overfitting as the validation loss fails to show a similar improvement.



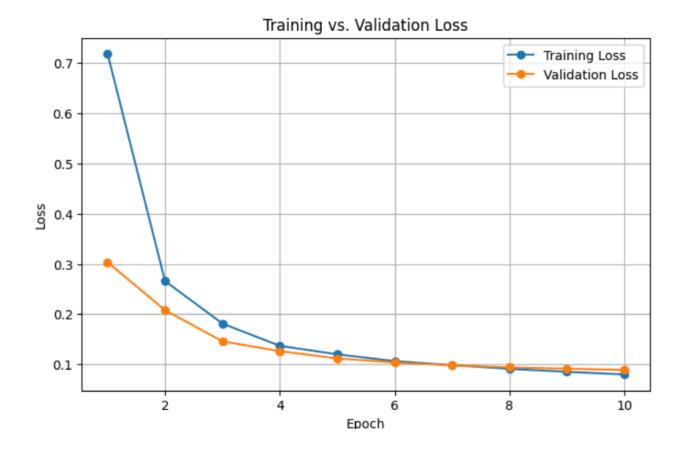
The graph shows encouraging signs of model improvement on the training data, as the training macro F1 score (blue line) increases with each epoch. However, the validation macro F1 score (orange line) stagnates, which suggests the model might be overfitting to the training data and failing to generalize well to unseen data. This graph visualizes the model's performance during training. While the training macro F1 score increases (blue line), indicating the model is learning the training data, the validation macro F1 score shows minimal improvement (orange line). This implies a risk of overfitting.

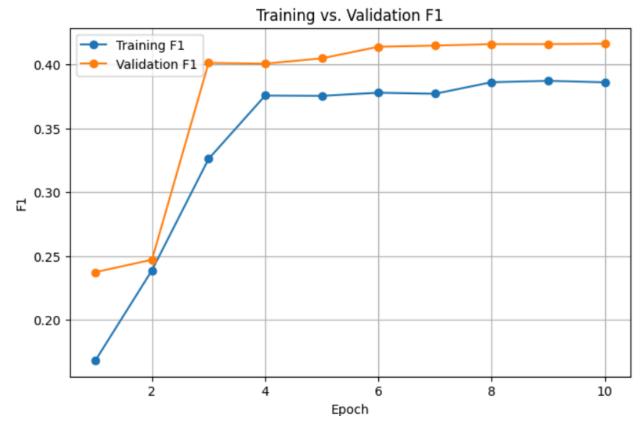
Loss: 0.12434762716293335 Metrics: [0.9681832194328308, 0.36756810545921326, 0.9711622595787048, 0.96146559715271]

0.4869181753317442

LSTM

F1 =

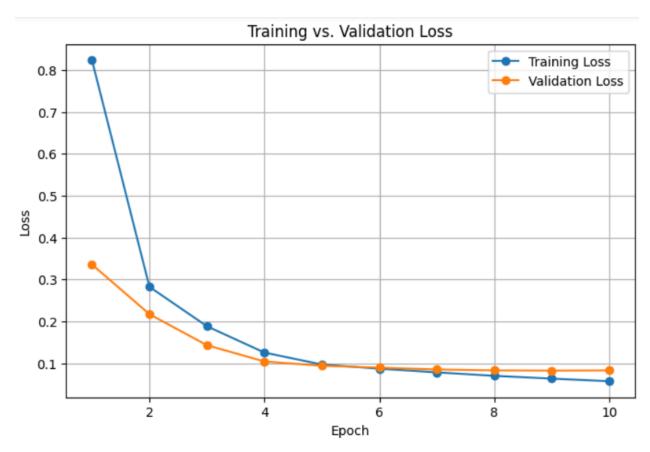


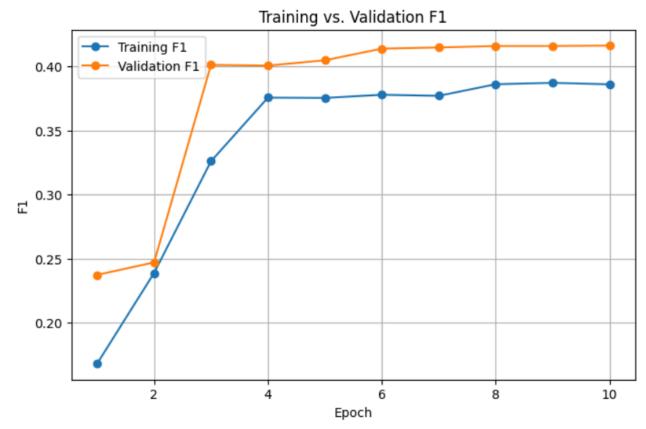


Loss: 0.11289666593074799
Metrics: [0.971768319606781, 0.3421818017959595, 0.9746668338775635, 0.9673296213150024]

 $F_1 = 0.47194246402119094$

GRU



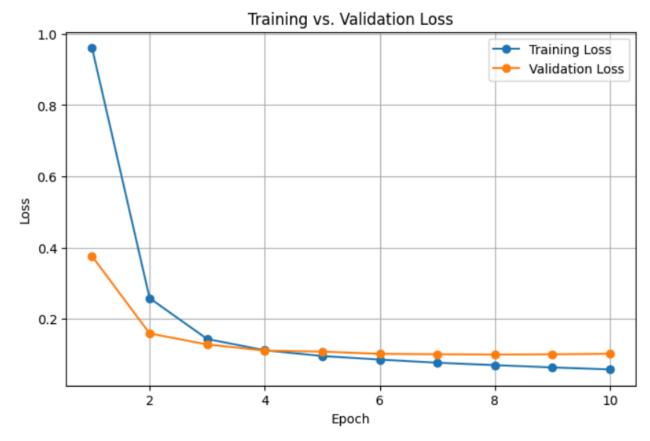


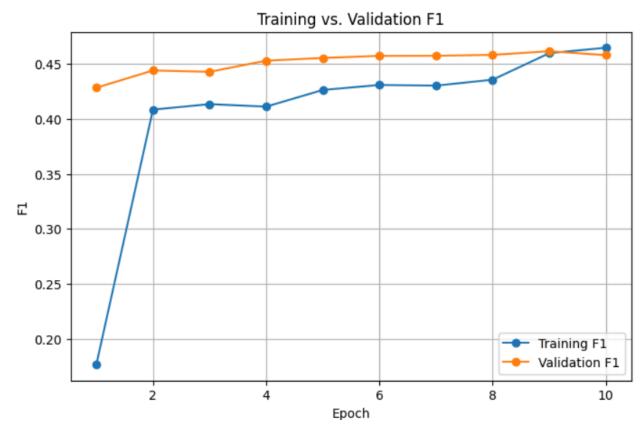
Loss: 0.11289666593074799 Metrics: [0.971768319606781, 0.3421818017959595, 0.9746668338775635, 0.9673296213150024]

$F_1 = 0.4885332363193045$

GloVe

Vanilla RNN

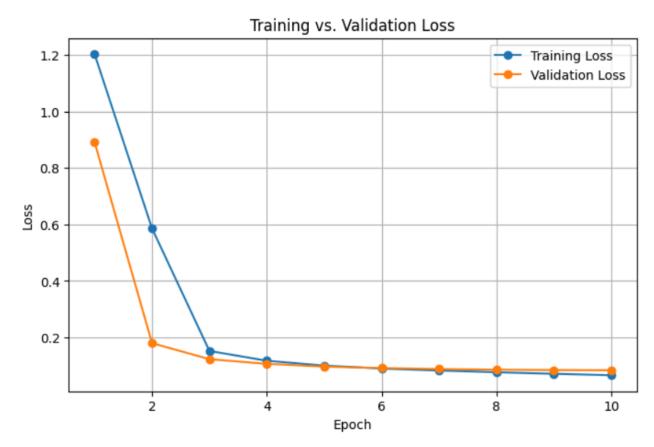


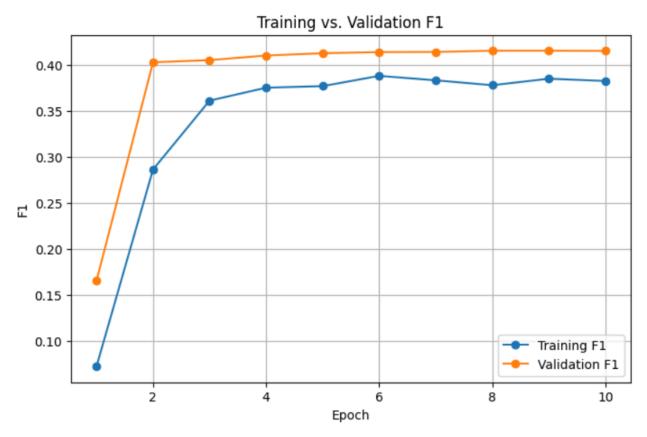


Loss: 0.18089483678340912 Metrics: [0.9433239698410034, 0.3489941358566284, 0.9559844732284546, 0.9364802837371826]

 $F_1 = 0.47186878375164776$

LSTM

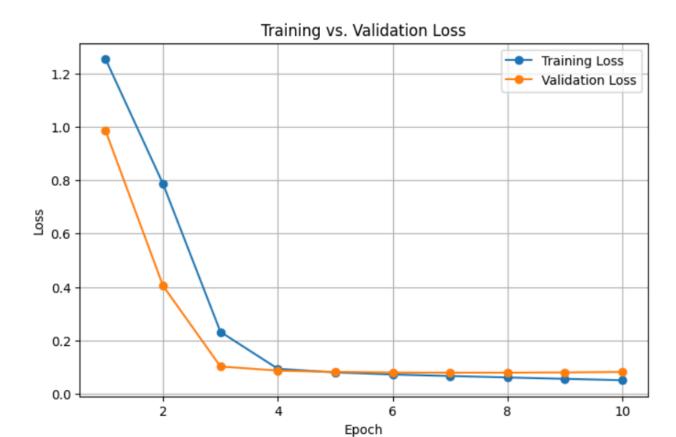


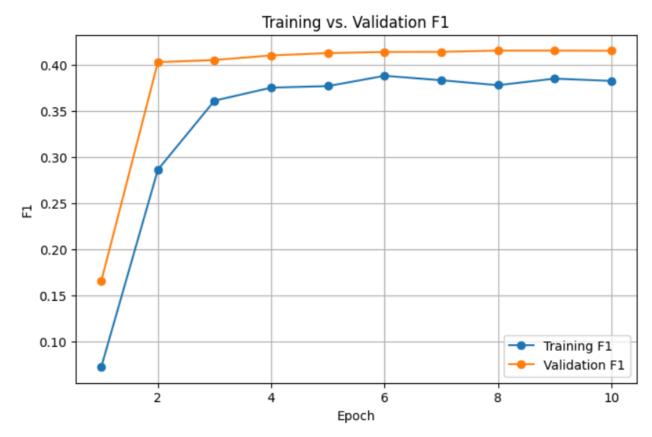


Loss: 0.1414303332567215 Metrics: [0.9557926654815674, 0.3272154629230499, 0.9658235311508179, 0.9504544734954834]

 $F_1 = 0.4670776793678769$

GRU



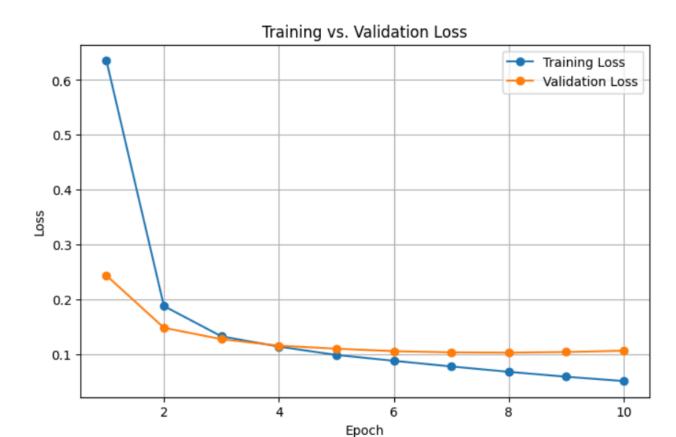


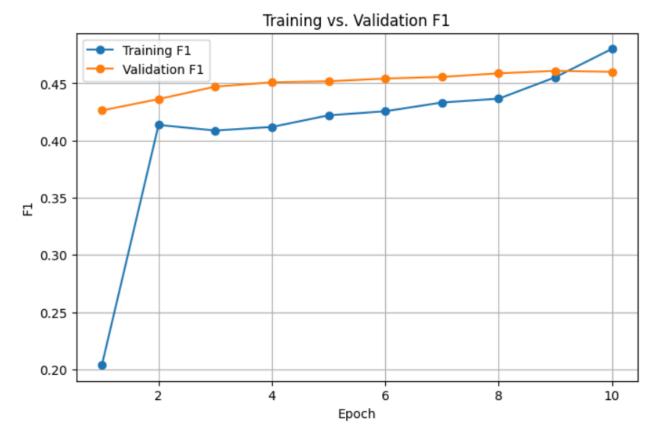
Loss: 0.1414303332567215 Metrics: [0.9557926654815674, 0.3272154629230499, 0.9658235311508179, 0.9504544734954834]

$F_{1} = 0.4749094833689165$

FastText

Vanilla RNN

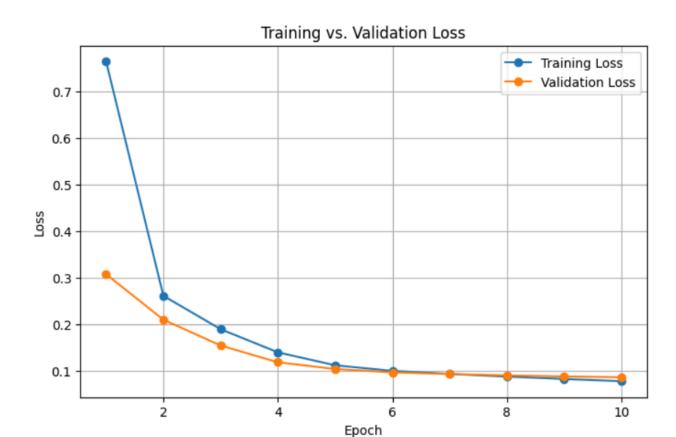


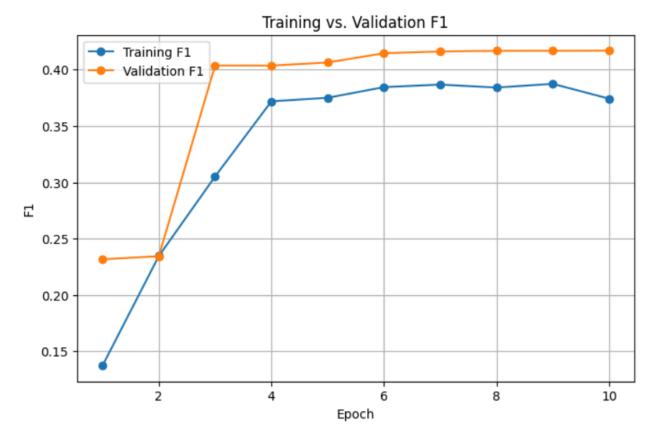


Loss: 0.13091148436069489 Metrics: [0.9618120789527893, 0.3651304543018341, 0.9679449200630188, 0.9545453786849976]

F1 = 0.4912289392945629

LSTM

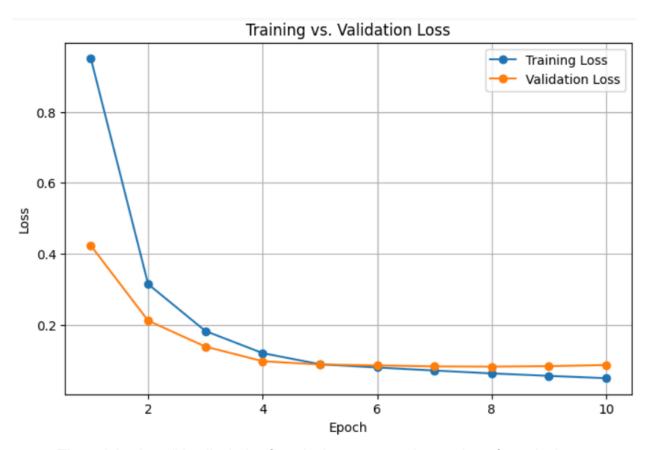


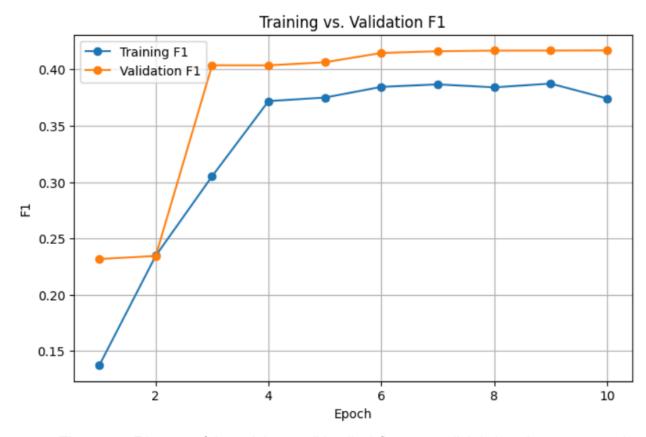


Loss: 0.10233476758003235 Metrics: [0.9702134132385254, 0.33980175852775574, 0.9750735759735107, 0.964545488357544]

F1 = 0.4738603646472841

GRU



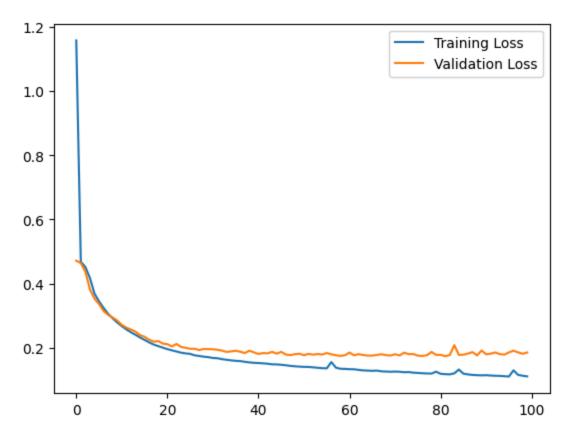


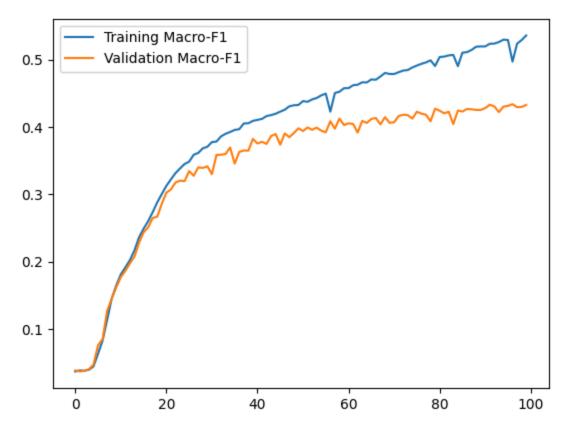
Loss: 0.10233476758003235 Metrics: [0.9702134132385254, 0.33980175852775574, 0.9750735759735107, 0.964545488357544]

F1 = 0.4905321483869017

Task 1:

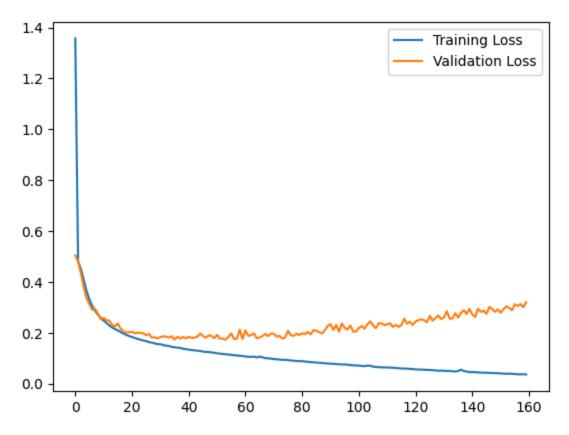
Word2Vect Vanilla RNN

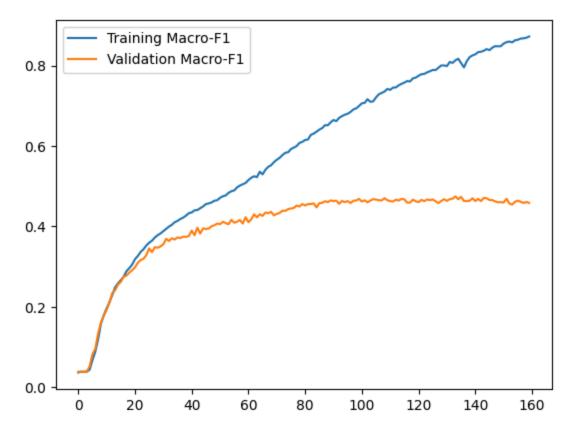




Test Loss: 0.1963, Test Macro-F1: 0.4264

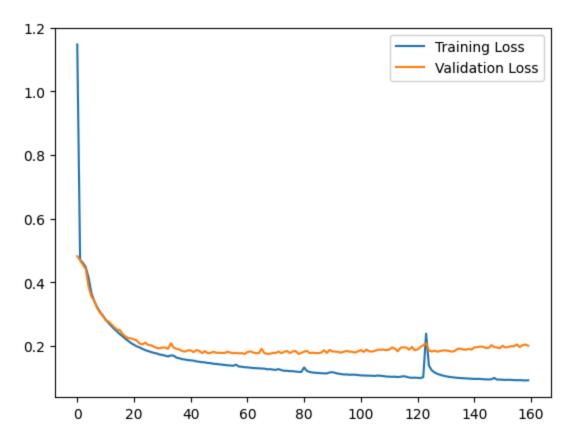
LSTM

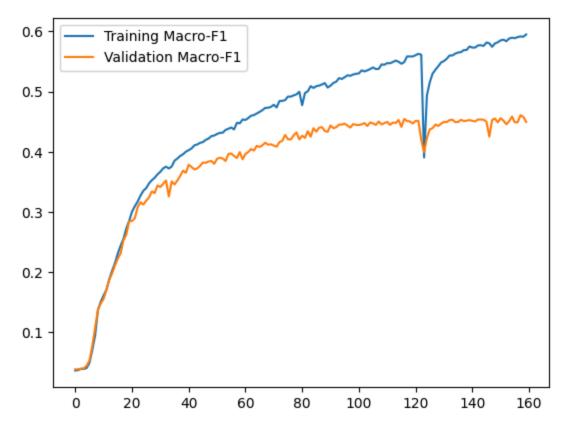




Test Loss: 0.3798, Test Macro-F1: 0.4197

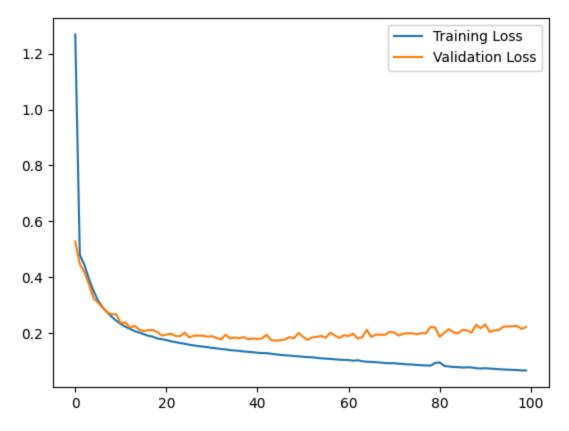
GRU

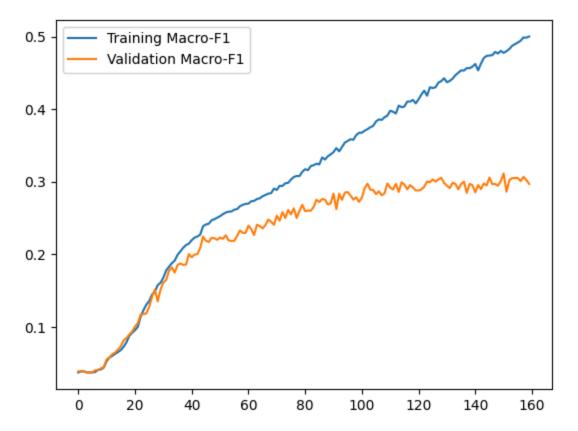




Test Loss: 0.2264, Test Macro-F1: 0.4264

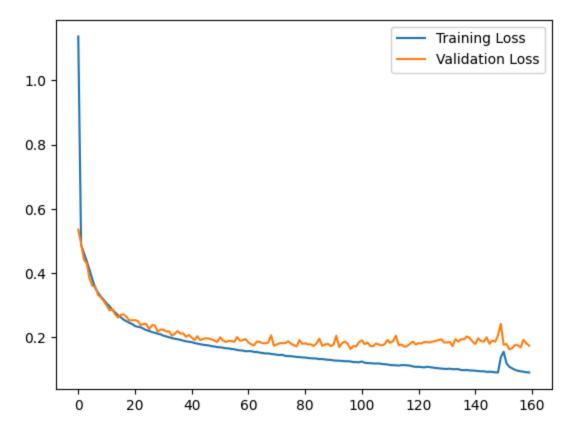
GloVe Vanilla RNN

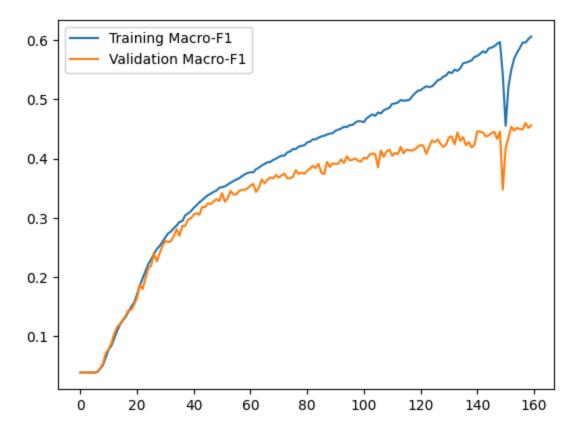




Test Loss: 0.3393, Test Macro-F1: 0.2949

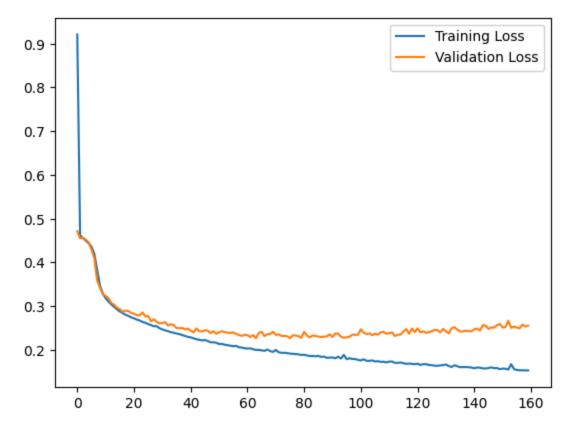
LSTM

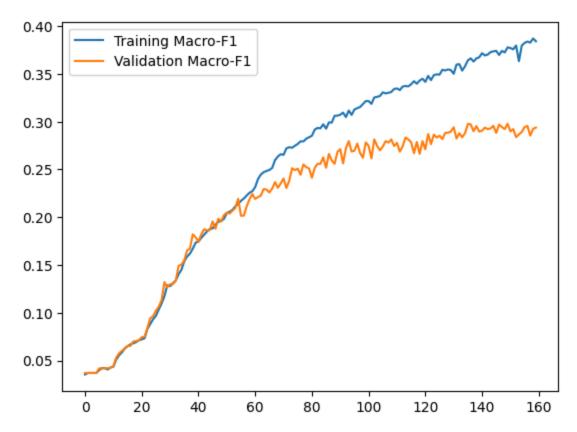




Test Loss: 0.1944, Test Macro-F1: 0.4642

GRU

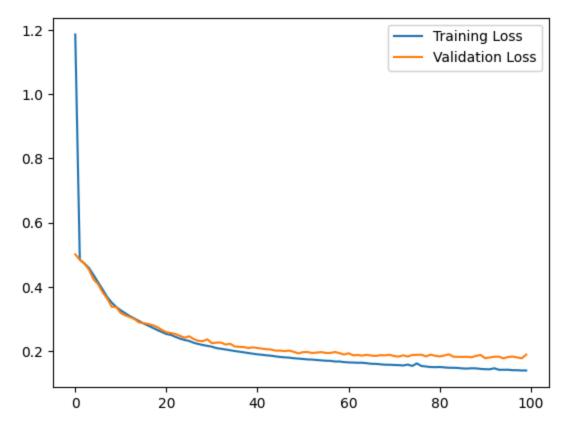


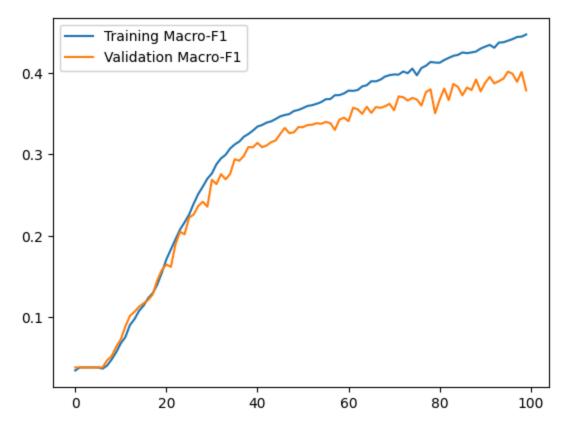


Test Loss: 0.2714, Test Macro-F1: 0.2813

Fasttext

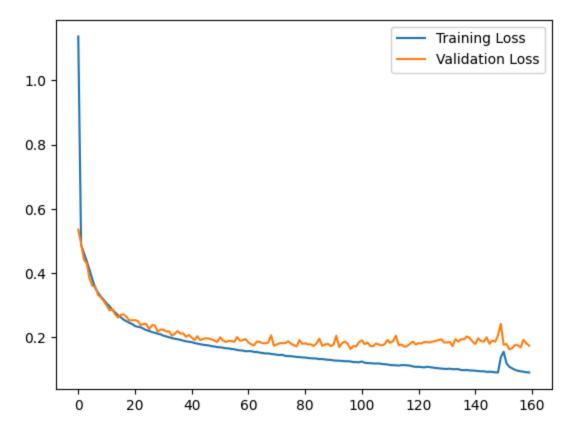
Vanilla RNN

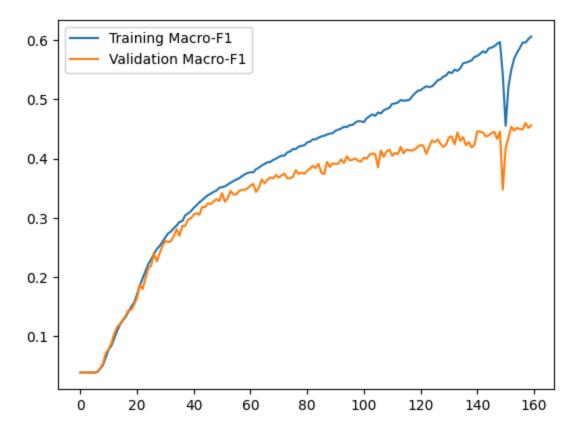




Test Loss: 0.2028, Test Macro-F1: 0.3813

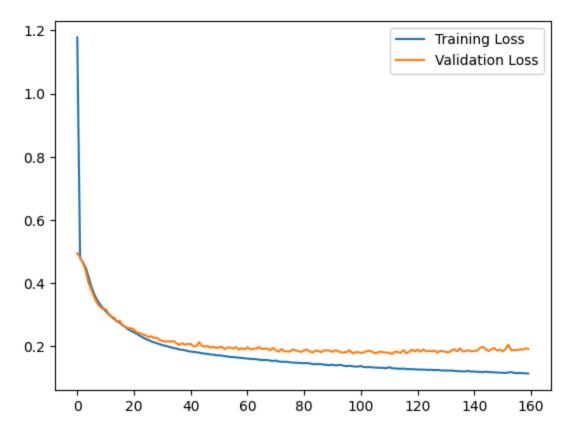
LSTM

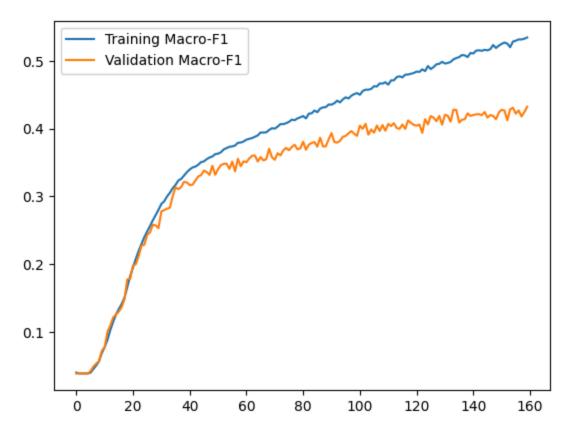




Test Loss: 0.1944, Test Macro-F1: 0.4642

GRU





Test Loss: 0.2107, Test Macro-F1: 0.4146

BiLSTM

NER Dataset

W2V:

The F1 score is 23.95%, the accuracy is 84.51%

FasttextL

F1 score is 10.28% and the accuracy is 85.28%

Glove:

The F1 score is 19.87% and the accuracy is 87.7%

These are all ran on 1epoch because it took too much time for it to run

Results for Task 2 in Part 2

Model	Word Embedding	Loss	Accuracy	Macro F1 Score
LSTM	Word2Vec	0.1194	0.9662	0.4822
RNN	Word2Vec	0.1185	0.9694	0.4973
GRU	Word2Vec	0.0889	0.9778	0.4841
LSTM	GloVe	0.1368	0.9555	0.4649
RNN	GloVe	0.1641	0.9470	0.4732
GRU	GloVe	0.1170	0.9665	0.4741
LSTM	FastText	0.1604	0.9645	0.4973
RNN	FastText	0.1183	0.9650	0.4846
GRU	FastText	0.0900	0.9765	0.4881

Results for Task 1 in Part 2

Model	Word Embedding	Test Loss	Test Macro-F1
RNN	Word2Vec	0.1963	0.4264
LSTM	Word2Vec	0.3798	0.4197
GRU	Word2Vec	0.2264	0.4264
RNN	GloVe	0.2514	0.2603
LSTM	GloVe	0.3393	0.2949
GRU	GloVe	0.2714	0.2813
RNN	FastText	0.2028	0.3813
LSTM	FastText	0.1944	0.4642
GRU	FastText	0.2184	0.4414

Individual Contribution:

- 1. Saumil Lakra(2021097): Part 1B[text processing], Models of ATE dataset in word2vec, the final compilation code for ATE and NER dataset.
- 2. Vishal Singh(20215): Embeddings of ATE and NER in Fasttext, GLOVE, graph plots, graph analysation, tables.
- 3. Sanskar Ranjan(2021096): Part 1A[text processing], BiLSTM CRF model.
- 4. Jeremiah Rokhum(2021533): BiLSTM CRF and NER dataset and other variations. It was fairly equal participation among all the members.