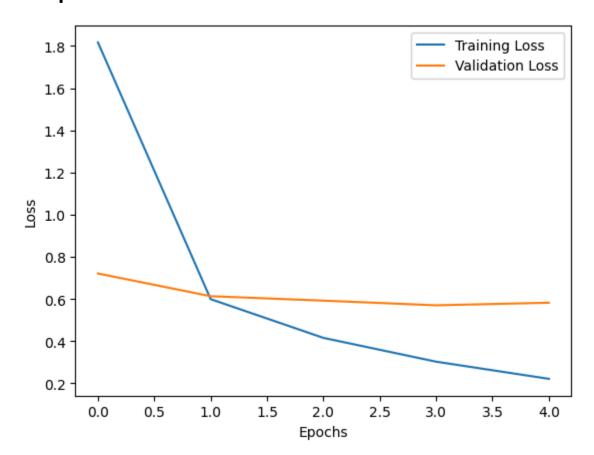
# **Group 36 – Assignment 3**

CSE 556: Natural Language Processing

# Task 1 - Text Similarity

# Setup 1A



Validation loss doesn't decrease after a few epochs, while training loss does.

Validation Pearson Correlation: 0.8676728865750375

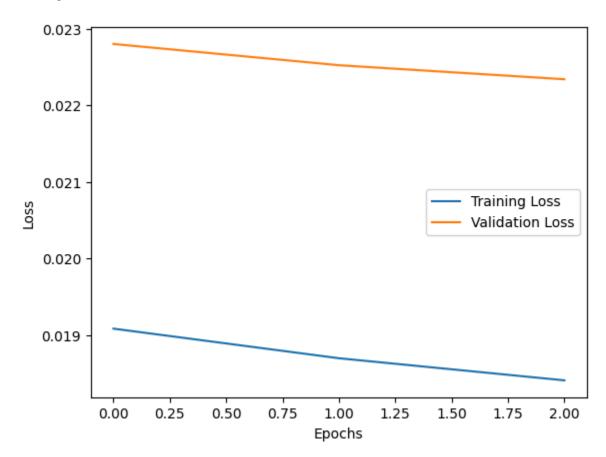
Test Pearson Correlation: 0.9784852902947985

# Setup 1B

Test-Set Pearson Correlation: 0.982463644410194

Validation-Set Pearson Correlation: 0.8631423846336786

# Setup 1C



Both training and validation loss doesn't decrease much and somewhat converge on fine-tuning. Validation Pearson Score: 0.8904479023415723

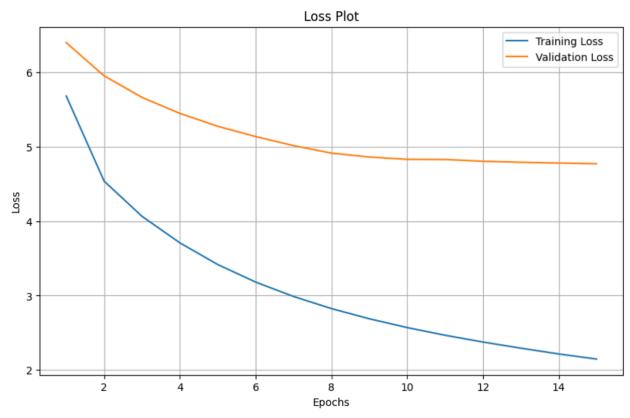
Test Pearson Score: 0.986926101342789

All the models performed similarly to some extent. Task1B and Task1C use the same model for the STS task. However, Task1C model provides better results on Pearson correlation due to fine-tuning.

## Task 2 - Machine Translation

### Setup 2A

We trained the model on 100K training data for 15 epochs. We noticed that validation loss decreases to a point then plateaus and starts to rise again while the training loss continues to go down. This is due to overfitting and hence we increased the training data to 100K, but the results improved only slightly. Also, we added weight decay to minimize the overfit. We employed early stopping and let the model train further past the plateau while retaining the best model.



Smooth loss and the expected trend in both training and validation, then validation loss plateaus

#### **Evaluation Score on Validation Set:**

BLEU-1: 100.0

BLEU-2: 71.42857142857143 BLEU-3: 30.76923076923077 BLEU-4: 8.333333333333334 BLEU score: 36.78763249927777

Average METEOR Score for validation data: 0.3043

Bert-Score for Validation data: Average Precision: 0.8538 Average Recall: 0.8632 Average F1: 0.8584

#### **Evaluation Score on Test Set:**

BLEU-1: 100.0 BLEU-2: 25.0

BLEU-3: 7.142857142857143 BLEU-4: 4.16666666666667 BLEU score: 16.515821590069034

Average METEOR Score for test data: 0.3153

Bert-Score for Test data: Average Precision: 0.8536 Average Recall: 0.8607 Average F1: 0.8570

### Setup 2B

#### **Evaluation Score on Validation Set:**

BLEU Score: 39.281465090051306

BLEU-1: 100.0 BLEU-2: 50.0

BLEU-3: 28.571428571428573 BLEU-4: 16.66666666666668

METEOR Score: 0.3697782590377402

**BERT Scores**:

Precision in BERT Score: 0.8268641829490662 Recall in BERT Score: 0.7662411332130432 F1 Score in BERT Score: 0.7942118644714355

#### **Evaluation Score on Test Set:**

BLEU Score: 0.0

BLEU-1: 66.666666666667

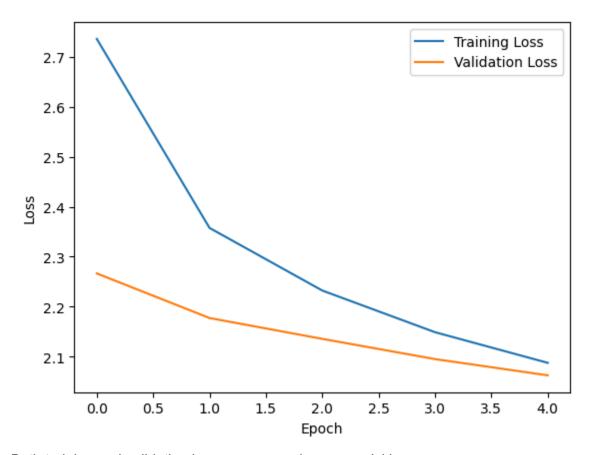
BLEU-2: 25.0 BLEU-3: 25.0 BLEU-4: 0.0

METEOR Score: 0.3889479564070389

**BERT Scores:** 

Precision in BERT Score: 0.8345420956611633 Recall in BERT Score: 0.7703850269317627 F1 Score in BERT Score: 0.8000085949897766

# Setup 2C



Both training and validation loss are converging very quickly.

### **Evaluation Score on Validation Set:**

BLEU Score: 11.478744233307168

BLEU-1 Score: 70.0

BLEU-2 Score: 11.11111111111111

BLEU-3 Score: 6.25

BLEU-4 Score: 3.5714285714285716 METEOR Score: 0.3227082170618216

**BERT Scores:** 

Precision in BERT Score: 0.79283207654953 Recall in BERT Score: 0.7551372647285461 F1 Score in BERT Score: 0.7728002071380615

#### **Evaluation Score on Test Set:**

BLEU Score: 0.0

BLEU-1 Score: 66.666666666667

BLEU-2 Score: 25.0 BLEU-3 Score: 25.0 BLEU-4 Score: 0.0

METEOR Score: 0.35205323785503273

BERT Scores:

Precision in BERT Score: 0.8002291321754456 Recall in BERT Score: 0.7628726363182068 F1 Score in BERT Score: 0.7803449034690857

We were able to finetune 't5-small' for German to English translation. The results were also good to a great extent. The translations visible in the Setup files are intuitive.

Results for Task 2A were also good but not as good as those of the fine-tuned model for Task 2C because t5-small is already pre-trained.

Task 2B model performed somewhat similar to the fine-tuned model as expected because it is the same model but on a different problem statement.

Results on the model for Task-2A are expected to increase provided that we use more training data, Transformers require a lot of data to perform well.

## **Contributions**

- 1. Khushdev Pandit:
  - a. helped in Task1A & Task1B
  - b. attempted Task1C, Task2B and Task2C
- Arjun Mehra: Task1A, Task1B
- 3. Pankaj: Helped in Task1C
- 4. Apurv Dube Task 2A