CS349 - ARTIFICIAL INTELLIGENCE II

ASSIGNMENT-4: ENCODER-DECODER

(Read all the instructions carefully & adhere to them.)

Date: 10 November 2023 Deadline: 17 November 2023

Total Credit: 20

Instructions:

1. The assignment should be completed and uploaded by 17 November 2023, 11:59 PM IST.

- **2.** Markings will be based on the correctness and soundness of the outputs. Marks will be deducted in case of plagiarism.
- **3.** Proper indentation and appropriate comments are mandatory.
- **4.** Make proper documentation of all results and observations with their analysis.
- **5.** You are supposed to make a group of at most three members.
- **6.** You should zip all the required files and name the zip file as roll_no_of_all_group_members .zip, e.g., 2101ai01_2102ai02_2101ai03.zip.
- 7. Upload your assignment (the zip file) in the google classroom with the class code: jtuplux
- **8.** Course Webpage: https://ai-nlp-ml-iitp.github.io/ai-cs349/
- **9.** For any queries regarding this assignment, you can contact:

Priyanshu Priya (priyanshu528priya@gmail.com) or

Manisha Boorja (manishaboorja@gmail.com)

Question

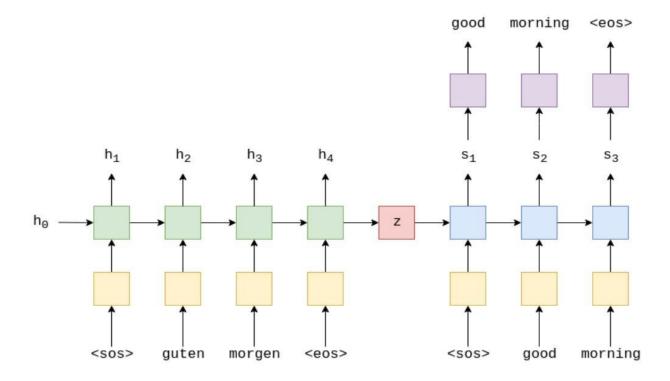
Problem Statement:

- The objective is to convert a English sentence to its Hindi counterpart using a Neural Machine Translation (NMT) system.
 - **Input**: Given Sentence in English. A start of the sentence (<sos>) and end of the sentence (<eos>) token needs to be appended.
 - "<sos>', 'it', 'is', 'raining', 'outside', '<eos>'
 - Output: Corresponding translated sentences in Hindi. A start of the sentence (<sos>) and end of the sentence (<eos>) token needs to be appended.
 - '<sos>', 'बाहर', 'वर्षा', 'हो', 'रही', 'है' '<eos>'
- You may consider the following details for the implementation.

- Input Vec(Wi input at the encoder): The word embeddings of the words from the input sentences will be the input to the model. You can use the Word2Vec or GLOVE embedding.
- Output Vec(Wo Input at the decoder): The word embeddings of the words from the input sentences will be the input to the model. You can use the Word2Vec or GLOVE embedding.
 - Link → Word2vec: http://vectors.nlpl.eu/repository/20/5.zip or https://drive.google.com/file/d/0B7XkCwpI5KDYNINUTTISS21pQmM/edit?usp=sharig
 - Link → Glove: http://nlp.stanford.edu/data/glove.840B.300d.zip
- Steps to use pre-trained word embeddings:
 - Prepare a dictionary of all the unique words in the dataset.
 - Load the word2vec or glove embeddings.
 - Get embeddings for each word and save them in a numpy or torch matrix.
- You may use any deep learning libraries such as TensorFlow, PyTorch, Keras etc. for the implementation. Use 300 dimensions for word embeddings.

• Neural Model

- o An LSTM based Encoder
- o An LSTM based Decoder



• **Dataset**: Download the dataset for Machine translation from here :

- https://drive.google.com/file/d/1jvZxoMsfVDvupZMqTMx11aHmMQFPyWG4/v iew?usp=sharing
- There are 3 files consisting of English and Hindi data
- Use the data in the files 'english.train.txt' and 'hindi.train.txt' for training. The sentences in the two files are aligned.
- Test your model using the files 'english.test.txt' and 'hindi.test.txt'
- Evaluation Metrics: Evaluate your model based on the following metrics:
 - BLEU score: BLEU looks at the overlap in the predicted and actual target sequences in terms of their n-grams. (Use the torchtext.data.metrics for computing bleu)
 - Using the gold samples from 'hindi.test.txt' compute the BLEU score.
- Loss Function: Use the CrossEntropyLoss function since it calculates both the log softmax as well as the negative log-likelihood for the predicted tokens.