## **Sentiment Analysis of IMDB Dataset**

#### 1. Preprocessing:

IMDB is a dataset of 25000 positive and 25000 negative review. I am performing preprocessing separately for training and testing but the procedure is almost same except tokenization.

- Collected data from IMDB dataset locally and saved into training and testing data.
- Removed HTML tags from review.
- Converted entire review into lowercase.
- Removed any string punctuation from review.
- Performed stemming on the review.
- Applied tokenization on data.
- Saved tokenization in training part, saved under model directory "tokenizer.pickle" and used the same file in testing.

### 2. Design Choice and Architecture for network:

- Performed deep learning convolutional model with RNN LSTM model on IMDB data to get the maximum accuracy.
- RNN model will just not help with building a model that consider the individual words but also remember the order they appear in which leads to a powerful model for sentiment prediction.
- I have used LSTM model as an RNN model as we shown above it is good tool for anything that has a sequence in it as the meaning of a word depends on the ones that preceded it.

#### • Architecture:

- **Embedding layer**: It is used to learn word embedding and perform NLP tasks such as text classification, sentiment analysis, etc.
  - Input dimension: 5000
    Output dimension: 32
  - o Input length: 500
- Dropout: Added 20% drop rate to prevent overfitting
- CNN layer:
  - Filter size: 32Activation: Relu
- Maxpooling: Added with pool size 2 which helps to extract complex features.
- **LSTM**: Added with 128 units.
- Dropout: Added 20% drop rate to prevent overfitting

## Output layer:

• Added dense layer as output layer with sigmoid activation function.

• **Optimizer**: Adam

• **Loss function**: binary\_crossentropy as we are having only two labels positive and negative.

# 3. Training and testing accuracy:

 Below training accuracy observed with 3 epochs as increasing more epochs leads to more time consumption and comprise in accuracy as well as increasing input dimension and input length in embedding layer result in less accuracy.

Training accuracy: 91.57%Testing accuracy: 87.79%