

## AML Lab 2 Report

**Team Number: 3**

**Team Name: The Anti Triads**

**Team Members:**

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### Dataset:

Our dataset contains 20,000 samples from the Kaggle-Quora Question Answer pairs dataset as downloaded using the web-service API provided for the team.

### Steps Followed:

- 1) The samples obtained from the web-service was converted into a pandas dataframe.
- 2) From the list of samples, all the question pairs were taken into a list and tokenized into a list of words.
- 3) All words were converted to lowercase and words below a frequency of 2 were removed to get a vocabulary size of 9048.
- 4) For all the words, the web service was invoked to get the corresponding embeddings and for words which threw an error when invoking the web service, the exception was caught, and a vector of 50 zeros was appended as the vector for those words.
- 5) The corresponding word-vector pair list obtained was converted into a pandas dataframe and an entry was added for out of vocabulary words corresponding to a vector of zeros.
- 6) Using the vector column of the dataframe and converting it into a numpy array, the embedding matrix was formed.
- 7) Using the indices for the words in the dataframe, each question from both the question pairs were converted into an embedding sequence and for words not present in the word-vector dataframe, the index corresponding to the out of vocabulary was appended into the sequence.
- 8) All sequences were then padded to the length of the maximum sequence (232) by appending all the sequences to the index corresponding to the vector of zeros.
- 9) The model was then designed by following an architecture similar to the diagram specified in the question-paper where each input was embedded separately, and an

LSTM was built for each of them, the output of which was concatenated and fed to a neural network with a single dense layer connected to an output layer with a sigmoid activation function. A detailed architecture will be provided in the next section.

- 10) The model was then trained on the sequences, using 17,000 sequence pairs for training (85% of the 20,000 samples) and the remaining 3000 sequence pairs for validation.
- 11) Numerous models with varying number of hidden units and other varying other hyperparameters were tried out, additionally adding regularizers and dropouts, while also adding additional dense layers to the neural network to see if there was any improvement in training, until the best network was finalized upon.

## Final Model Architecture

Layer (type)	Output Shape	Param #	Connected to
input_5 (InputLayer)	(None, 232)	0	
input_6 (InputLayer)	(None, 232)	0	
embedding_6 (Embedding)	(None, 232, 50)	452800	input_5[0][0] input_6[0][0]
lstm_16 (LSTM)	(None, 20)	1120	embedding_6[0][0] embedding_6[1][0]
concatenate_11 (Concatenate)	(None, 10)	0	lstm_16[0][0] lstm_16[1][0]
dropout_20 (Dropout)	(None, 10)	0	concatenate_11[0][0]
batch_normalization_18 (Batch Normalization)	(None, 10)	40	dropout_20[0][0]
dense_19 (Dense)	(None, 20)	220	batch_normalization_18[0][0]
dropout_21 (Dropout)	(None, 20)	0	dense_19[0][0]

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batch_normalization_19	(BatchNo (None, 20)	80	dropout_21[0][0]
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dense_20 (Dense)	(None, 1)	21	
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batch\_normalization\_19[0][0]

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Total params: 454,281  
Trainable params: 1,421  
Non-trainable params: 452,860

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### More model details:

Optimizer: Adadelat optimizer.

Batch size: 64

Learning Rate left as default as suggested by Keras documentation for the Adadelat optimizer.

Number of epochs: 25 (training time was very high on a CPU machine)

LSTM: 20 units

Dense layer:

1. Number of units: 20
2. Activation function: Exponential Linear Unit
3. Regularizers: Kernel Regularizer: l2(0.1)
4. Activity Regularizer: l1(0.01)
5. Dropout 0.2

Loss function: Mean squared error

The regularizers, dropouts and batch normalization were added to the network to help deal with both overfitting and to help in accelerating convergence to the best possible minima.

### Model Results:

Training Accuracy: **0.6264**

Training Loss: 0.3628

Validation Accuracy: **0.6367**

Validation Loss: 0.3646