

CODE SUMMARY

For this course's final project, I am building a Chatbot. Here is a summary of my results. Here are the file names and their objectives:

- Parser_1: Extracts data from What's-app log and creates a csv
- Process_whatsapp_data_2: Builds padded input sequences (question, answer)
- Train_bot_3: training takes place in this file
- Test_bot_4: testing takes place in this file

Dataset used: [Movie dialogue and WhatsApp log](#)

To view text files in formatted space use Sublime.txt or Notepad++. I am still trying to improve the context of the bot replies.

EXPLANATION: Here I am extracting WhatsApp conversation logs from my WhatsApp chat history. WhatsApp has a feature that allows us to view all chat history in text format. After processing this data, I trained my model on this data.

```
# -----< Read Whatsapp chatlog >-----
def open_file(self):
    arq = codecs.open(self.inputFileName, "r", "utf-8-sig")
    content = arq.read()
    arq.close()
    lines = content.split("\n")
    lines = [l for l in lines if len(l) > 4]
    for l in lines:
        self.raw_messages.append(l.encode("utf-8"))
```

After obtaining our dataset we perform padding which pre-fixes our input sequence length.

```
# -----< Padding the question and answer / input and output generated above >-----
def pad_question_answers(self):
    print ("Reading the context data...")
    q = open(self.questions_file, 'r')
    questions = q.read()
    print ("Reading the answer data...")
    a = open(self.answers_file, 'r')
    answers = a.read()
    all = answers + questions
    print ("Tokenazing the answers...")
    paragraphs_a = [p for p in answers.split('\n')]
    paragraphs_b = [p for p in all.split('\n')]
    paragraphs_a = ['BOS ' + p + ' EOS' for p in paragraphs_a]
    paragraphs_b = ['BOS ' + p + ' EOS' for p in paragraphs_b]
    paragraphs_b = ' '.join(paragraphs_b)
    tokenized_text = paragraphs_b.split()
    paragraphs_q = [p for p in questions.split('\n')]
    tokenized_answers = [p.split() for p in paragraphs_a]
    tokenized_questions = [p.split() for p in paragraphs_q]

    vocab = pickle.load(open(self.vocabulary_file, 'rb'))

    index_to_word = [x[0] for x in vocab]
    index_to_word.append(self.unknown_token)
    word_to_index = dict([(w,i) for i,w in enumerate(index_to_word)])

    print ("Using vocabulary of size %d." % self.vocabulary_size)
    print ("The least frequent word in our vocabulary is '%s' and appeared %d times." % (vocab[-1][0], vocab[-1][1]))
```

OUR MODEL:

This image demonstrates how easy it is to design our model using keras and then perform optimizations on it. I have used SGD (Stochastic gradient descent) as the optimizer.

```
# -----<Build our model for training using SGD optimizer>-----
def build_model(self, embedding_matrix):
    gradDescent = SGD(lr=0.00005)
    input_context = Input(shape=(self.maxlen_input,), dtype='int32', name='input_context')
    # input_context = tf.cast(input_context,tf.int32)
    input_answer = Input(shape=(self.maxlen_input,), dtype='int32', name='input_answer')
    # input_answer = tf.cast(input_answer,tf.int32)
    LSTM_encoder = LSTM(self.sentence_embedding_size, init= 'lecun_uniform')
    LSTM_decoder = LSTM(self.sentence_embedding_size, init= 'lecun_uniform')
    if os.path.isfile(self.weights_file):
        Shared_Embedding = Embedding(output_dim=self.word_embedding_size, input_dim=self.dictionary_size, input_length=self.maxl
    else:
        Shared_Embedding = Embedding(output_dim=self.word_embedding_size, input_dim=self.dictionary_size, weights=[embedding_mat

    word_embedding_context = Shared_Embedding(input_context)
    context_embedding = LSTM_encoder(word_embedding_context)
    word_embedding_answer = Shared_Embedding(input_answer)
    answer_embedding = LSTM_decoder(word_embedding_answer)
    merge_layer = merge([context_embedding, answer_embedding], mode='concat', concat_axis=1)
    print('dsize',self.dictionary_size/2)

    out = Dense(int(self.dictionary_size/2), activation="relu")(merge_layer)
    out = Dense(int(self.dictionary_size), activation="softmax")(out)

    model = Model(input=[input_context, input_answer], output = [out])
    model.compile(loss='categorical_crossentropy', optimizer=gradDescent)

    if os.path.isfile(self.weights_file):
        model.load_weights(self.weights_file)

    return model
```

Here I am making use of a pre-trained model to lower my computational effort and results. These are nothing but weights which are embedded and then used in my own model instead of using random values.

```
# -----<Apply transfer learning using glove directory >-----
# source for glove dir :
# https://nlp.stanford.edu/projects/glove/
# glove.6b.zip
def transfer_learning(self):
    f = open(os.path.join(self.GLOVE_DIR, 'glove.6B.100d.txt'),encoding="utf8")
    for line in f:
        values = line.split()
        word = values[0]
        coefs = np.asarray(values[1:], dtype='float32')
        self.embeddings_index[word] = coefs
    f.close()

    print('Found %s word vectors.' % len(self.embeddings_index))
    embedding_matrix = np.zeros((self.dictionary_size, self.word_embedding_size))

    # Loading our vocabulary:
    vocabulary = pickle.load(open(self.vocabulary_file, 'rb'))

    # Using the Glove embedding:
    i = 0
    for word in vocabulary:
        embedding_vector = self.embeddings_index.get(word[0])

        if embedding_vector is not None:
            # words not found in embedding index will be all-zeros.
            embedding_matrix[i] = embedding_vector
            i += 1

    return embedding_matrix
```

RESULTS:

This is how we test our model by using the `model.predict` inbuilt function available in keras. The arguments include the input parameters. Most probable result is printed, and probability is also displayed as part of result.

```
# -----< Testing our model using inbuilt predict function in keras >-----
def predict(self, input):
    flag = 0
    probability = 1
    ans_partial = np.zeros((1, self.maxlen_input))
    ans_partial[0, -1] = 2 # the index of the symbol BOS (begin of sentence)
    for k in range(self.maxlen_input - 1):
        ye = self.model.predict([input, ans_partial])
        yel = ye[0, :]
        p = np.max(yel)
        mp = np.argmax(yel)
        ans_partial[0, 0:-1] = ans_partial[0, 1:]
        ans_partial[0, -1] = mp
        if mp == 3: # the index of the symbol EOS (end of sentence)
            flag = 1
        if flag == 0:
            probability = probability * p
    text = ''
    for k in ans_partial[0]:
        k = k.astype(int)
        if k < (self.dictionary_size-2):
            w = self.vocabulary[k]
            text = text + w[0] + ' '
    return(text, probability)
```

RESULT ANALYSIS:

Train log: Full log can be found in the Traininglog.txt file

As we can see the model is training and the error value is decreasing. The text below each epoch is one instance of training input

Test log result analysis:

- Artificial Neural Network – Assignment 2

```

Welcome to chatbot demo:

computer: hi , Kunal ! My name is RedEye.
user: Good to know!
computer: what do you mean ?      (with probability of 0.088123)
user: I mean i am happy
computer: me too .      (with probability of 0.165586)
user: That's nice
computer: it is worse than that .      (with probability of 0.047418)
user: What are you talking about?
computer: what is the matter ?      (with probability of 0.304714)
user: Do you want to go to a movie
computer: that sounds like a good idea .      (with probability of 0.335707)
user: I will book ticket for tomorrow
computer: what is the matter ?      (with probability of 0.185682)
user: It seems i dont have money for both of us
computer: you are right .      (with probability of 0.167489)
user: Can you lend me some ?
computer: sure . i will have the too in .      (with probability of 0.000033)
user: Get me a job
computer: what was your last job ?      (with probability of 0.857101)
user: Doctor
computer: what happened ?      (with probability of 0.926905)
user: I left
computer: i am sorry for you .      (with probability of 0.073566)
user: Thanks
computer: you are welcome .      (with probability of 0.159442)
user: Ok lets switch topics
computer: i do not want to talk about it .      (with probability of 0.167818)
user: Do you like Computer Science?
computer: yes , i do . i am learning english and literature .      (with probability of 0.000969)
user: Yes its the same thing
computer: what happened ?      (with probability of 0.116010)
user: Do you want me to teach you French?
computer: i am interested in second class .      (with probability of 0.029071)
user: Let me know what time will be the best

```

BUILD INSTRUCTIONS:

First install Anaconda to set up the build environment.

I used python version 3.5 since tensor flow support was not there for certain libraries. Here are the build commands for Linux.

Required files: Vocabulary_file.txt, data.txt (Movie Data), Traininglog.txt (WhatsApp data), [Glove6B](#) (Transfer learning)

```

conda create --name myp35 python=3.5
python Parser_1.py
pip install 'install dependencies here'
python Process_WhatsAppData.py
pip install 'install dependencies here'
python Train_bot.py
pip install 'install dependencies here'
python Test_bot.py
pip install 'install dependencies here'

```