

###MACHINE LEARNING ASSIGNMENT Grammatical Error Correction

Grammatical error correction is the task of automatically correcting grammatical errors in a text.

A grammatical error correction system takes an erroneous sentence as input and is expected to find all the above errors to transform the sentence into the corrected version. For example –

- Incorrect Sentence 1: “She see Tom is caught by policeman in park at last night.”
- Corrects Sentence 1: “She saw Tom caught by a policeman in the park last night.”
- Incorrect Sentence 2: “It is not of only your business.”
- Corrects Sentence 2: “It is none of your business.”

In [48]:

```

1  ! pip install jsonlines
2  import pandas as pd
3  import numpy as np
4  import os
5  import jsonlines
6  import re
7  import matplotlib.pyplot as plt
8  import seaborn as sns
9  import warnings
10 warnings.filterwarnings('ignore')
11 from tqdm import tqdm
12 from nltk.corpus import stopwords
13 from wordcloud import WordCloud
14 import nltk
15 from nltk.tokenize import word_tokenize
16 from collections import Counter
17 from tqdm import tqdm
18 import tensorflow as tf
19 from tensorflow.keras.preprocessing.sequence import pad_sequences
20 from sklearn.model_selection import train_test_split
21 from tqdm import tqdm
22 from tensorflow.keras.preprocessing.text import Tokenizer
23 from tensorflow.keras import layers
24 import nltk.translate.bleu_score as bleu
25 from tensorflow.keras import Model

```

Requirement already satisfied: jsonlines in /usr/local/lib/python3.7/dist-packages (3.0.0)

Requirement already satisfied: attrs>=19.2.0 in /usr/local/lib/python3.7/dist-packages (from jsonlines) (21.4.0)

Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages (from jsonlines) (3.10.0.2)

###Dataset Used

1. Lang-8
2. NUS Social Media Text Normalization and Translation Corpus

We would be combining both the datasets to train out model.

####Loading the dataset from my google drive.

In [49]:

```
1 f1 = open("/content/drive/MyDrive/DATASET/entries.train")
2 lines1 = f1.readlines()
3 inp1 = []
4 tgt1 = []
5
6 for i in lines1:
7     lst = i.split("\t")
8
9
10     if len(lst)>5 :
11         inp1.append(lst[-2])
12         tgt1.append(lst[-1])
```

In []:

```
1 '''THIS CELL READS THE SMS_TEXT DATASET FILE AND EXTRACTS INCORRECT AND CORRECT SENTENCES'''
2
3
4 f2 = open("/content/drive/MyDrive/DATASET/en2cn-2k.en2nen2cn", "r", encoding="UTF-8") # f
5
6 lines2 = f2.readlines()
7 inp2 = []
8 tgt2 = []
9
10
11 for i in range(2000):
12     inp2.append(lines2[i*3])
13     tgt2.append(lines2[i*3+1])
```

In []:

```
1 '''Combining two datasets'''
2
3 df = pd.DataFrame()
4 df["input"] = inp1+inp2
5 df["output"] = tgt1+tgt2
6 df["y"] = list("1"*len(inp1)) + list("2"*len(inp2))
```

In []:

```
1 df
```

This function removes the space between the words also between the sentences

```
ca n't ==> can't
I 'm ==> I'm ...etc
```

In []:

```

1
2 def remove_spaces(text):
3     text = re.sub(r" '(\w)", r"' \1", text)
4     text = re.sub(r" \,", ",", text)
5     text = re.sub(r" \.", ".", text)
6     text = re.sub(r" \!", "!", text)
7     text = re.sub(r" \?", "?", text)
8     text = re.sub(r" n't", "n't", text)
9     text = re.sub(r"[\(\)\;\_\^`\/]", "", text)
10
11     return text
12
13
14 '''THIS FUNCTION DECONTRACTS THE CONTRACTED WORDS'''
15 #REF : https://stackoverflow.com/questions/19790188/expanding-english-language-contract
16
17 def decontract(text):
18     text = re.sub(r"won't", "will not", text)
19     text = re.sub(r"can't", "can not", text)
20     text = re.sub(r"n't", "not", text)
21     text = re.sub(r"\ 're", " are", text)
22     text = re.sub(r"\ 's", " is", text)
23     text = re.sub(r"\ 'd", " would", text)
24     text = re.sub(r"\ 'll", " will", text)
25     text = re.sub(r"\ 't", " not", text)
26     text = re.sub(r"\ 've", " have", text)
27     text = re.sub(r"\ 'm", " am", text)
28     return text
29
30
31 '''THIS FUNCTION PREPROCESSES THE TEXT '''
32 def preprocess(text):
33     text = re.sub(r"\n", "", text)
34     text = remove_spaces(text) # REMOVING UNWANTED SPACES
35     text = re.sub(r" \.", ".", text)
36     text = re.sub(r" \!", "!", text)
37     text = decontract(text) # DECONTRACTION
38     text = re.sub(r"^[A-Za-z0-9 ]+", "", text)
39     text = text.lower()
40     return text

```

In []:

```

1 '''HERE WE ARE APPLYIN PREPROCESS FUNCTION TO INPUT AND OUTPUT SENTENCES'''
2
3 df["processed_input"] = df.input.apply(preprocess)
4 df["processed_output"] = df.output.apply(preprocess)

```

In []:

```
1 dataset=df.drop(['input', 'output'],axis=1)
```

In []:

```
1 dataset
```

In []:

```
1 '''CHECK FOR NULL VALUES'''
2
3 dataset.info()
```

In []:

```
1 '''THIS CELL REMOVES ROWS WITH NULL VALUES IN INPUT AND OUTPUT TEXT'''
2
3 dataset = dataset[dataset.processed_input.notnull()]
4 dataset = dataset[dataset.processed_output.notnull()]
```

In []:

```
1 dataset = dataset.drop_duplicates()
```

In []:

```
1 dataset
```

In []:

```
1 dataset.info()
```

Data is preprocdd and stored in to the drive with .csv format

In []:

```
1 dataset[["processed_input", "processed_output", "y"]].to_csv("/content/drive/MyDrive/DATA/processed_data.csv")
```

###Here in this program I have used Attention Mechanism for grammer correction

#####Attention Mechanism Attention Mechanism is a very ingenious idea in Machine Learning which clones the humanistic way of grasping information.It can be implemented using the idea of Recurrent NN

In []:

```
1
2 dataset= pd.read_csv("/content/drive/MyDrive/DATASET/processed_data.csv")
3 dataset.columns = ["enc_input", "dec_input", "y"]
4 dataset["dec_output"] = dataset.dec_input
5 dataset
```

Adding start and end to the o/p sentence for the usefull ness of the decoder

In []:

```
1
2 dataset["dec_input"] = "<start> " + dataset["dec_input"]
3 dataset["dec_output"] = dataset["dec_output"] + " <end>"
4 dataset
```

Splitting And Sampling around 100k datapoints

In []:

```
1 df_sampled = pd.concat((dataset[dataset.y==1].sample(frac= 0.2,random_state=1),dataset[dataset.y==0].sample(frac=0.2,random_state=1)))
```

In []:

```
1 ## Splitting the data .
2 df_train ,df_val = train_test_split(df_sampled,test_size=0.2,random_state = 3, stratify=dataset.y)
```

In []:

```
1
2 df_train["dec_input"].iloc[0] = df_train.iloc[0]["dec_input"] + " <end>"
3 df_train
```

In [54]:

```
1
2 df_val
```

Out[54]:

	enc_input	dec_input	y	dec_output
71342	in the last stage add the mixture of potato an...	<start> in the last stage add the mixture of p...	1	in the last stage add the mixture of potato an...
395034	of course the experiences of japanese are also...	<start> of course the experiences of the japan...	1	of course the experiences of the japanese are ...
208880	but watching other is play gives me many stimu...	<start> but watching others play is exciting a...	1	but watching others play is exciting after all...
198826	and this is my new tattoos	<start> and these are my new tattoos	1	and these are my new tattoos <end>
313016	i write a diary in english for the first time	<start> i am writing a diary in english for th...	1	i am writing a diary in english for the first ...
...
247300	i love family	<start> i love my family	1	i love my family <end>
505055	ok i going soon and also send xyan home at the...	<start> ok i am going soon and also send xyan ...	2	ok i am going soon and also send xyan home at ...
126924	house car people had been washed away by the b...	<start> houses cars and people had been washed...	1	houses cars and people had been washed away by...
53162	on the weekend i had to go back to my home to ...	<start> on the weekend i had to go back to my ...	1	on the weekend i had to go back to my home to ...
10995	she lilves in us and studys to be nurse	<start> she lives in us and is studying to be ...	1	she lives in us and is studying to be a nurse ...

20556 rows × 4 columns

In []:

```
1 ## Here 1000 points are uses from the dataframe as test data.
2 np.random.seed(5)
3 df_test = dataset.loc[np.random.choice(np.array([x for x in dataset.index.values if x r
4 df_test
```

In []:

```

1  ## Tokenizer for encoder I/P
2  tk_inp = Tokenizer()
3  tk_inp.fit_on_texts(df_train.enc_input.apply(str))

```

In []:

```

1  ## Tokenizer for decoder I/P
2  tk_out = Tokenizer(filters='! "$%&()*+,-./:;=?@[\\]^_`{|}~\t\n' )
3  tk_out.fit_on_texts(df_train.dec_input.apply(str))

```

This class converts the text data to int seq. and returns the padded seq.

In []:

```

1
2  class Dataset :
3      def __init__(self, data , tk_inp ,tk_out, max_len):
4
5          self.encoder_inp = data["enc_input"].apply(str).values
6          self.decoder_inp = data["dec_input"].apply(str).values
7          self.decoder_out = data["dec_output"].apply(str).values
8          self.tk_inp = tk_inp
9          self.tk_out = tk_out
10         self.max_len = max_len
11
12         def __getitem__(self,i):
13
14             self.encoder_seq = self.tk_inp.texts_to_sequences([self.encoder_inp[i]])
15             self.decoder_inp_seq = self.tk_out.texts_to_sequences([self.decoder_inp[i]])
16             self.decoder_out_seq = self.tk_out.texts_to_sequences([self.decoder_out[i]])
17             self.encoder_seq = pad_sequences(self.encoder_seq, padding="post",maxlen = self.max_len)
18             self.decoder_inp_seq = pad_sequences(self.decoder_inp_seq, padding="post",maxlen = self.max_len)
19             self.decoder_out_seq = pad_sequences(self.decoder_out_seq ,padding="post", maxlen = self.max_len)
20             return self.encoder_seq , self.decoder_inp_seq, self.decoder_out_seq
21
22         def __len__(self):
23             return len(self.encoder_inp)

```

Covertng data into batches

In []:

```
1 class Dataloader(tf.keras.utils.Sequence):
2     def __init__(self, batch_size, dataset):
3
4         self.dataset = dataset
5         self.batch_size = batch_size
6         self.totl_points = self.dataset.encoder_inp.shape[0]
7
8     def __getitem__(self, i):
9
10        start = i * self.batch_size
11        stop = (i+1)*self.batch_size
12        batch_enc = []
13        batch_dec_input = []
14        batch_dec_out = []
15
16        for j in range(start, stop):
17
18            a, b, c = self.dataset[j]
19            batch_enc.append(a[0])
20            batch_dec_input.append(b[0])
21            batch_dec_out.append(c[0])
22
23        batch_enc = (np.array(batch_enc))
24        batch_dec_input = np.array(batch_dec_input)
25        batch_dec_out = np.array(batch_dec_out)
26
27        return [batch_enc , batch_dec_input], batch_dec_out
28
29    def __len__(self):
30        return int(self.totl_points/self.batch_size)
```

In []:

```
1 train_dataset = Dataset(df_train, tk_inp, tk_out, 35)
2 train_dataloader = Dataloader( batch_size = 32, dataset=train_dataset)
3
4 val_dataset = Dataset(df_val , tk_inp, tk_out, 35)
5 val_dataloader = Dataloader(batch_size=32 , dataset=val_dataset)
```

Encoder Class

In []:

```

1  ## DEFINING THE ENCODER LAYER AS A FUNCTION
2  class Encoder(tf.keras.layers.Layer):
3
4      def __init__(self, vocab_size, emb_dims, enc_units, input_length, batch_size):
5          super().__init__()
6
7          self.batch_size = batch_size
8          self.enc_units = enc_units
9          self.embedding = layers.Embedding(vocab_size, emb_dims)
10
11         self.lstm = layers.LSTM(self.enc_units, return_state=True, return_sequences=True)
12
13     def call(self, enc_input, states):
14         emb = self.embedding(enc_input)
15         enc_output, state_h, state_c = self.lstm(emb, initial_state=states)
16         return enc_output, state_h, state_c
17     def initialize(self, batch_size):
18
19         return tf.zeros(shape=(batch_size, self.enc_units)), tf.zeros(shape=(batch_size, self.enc_units))

```

###Attention Mechanism

In []:

```

1  class Attention(tf.keras.layers.Layer):
2
3      def __init__(self, units):
4          super().__init__()
5
6          self.dense = layers.Dense(units)
7
8      def call(self, enc_output, dec_state):
9
10         dec_state = tf.expand_dims(dec_state, axis=-1)
11         dense_output = self.dense(enc_output)
12         score = tf.matmul(dense_output, dec_state)
13         att_weights = tf.nn.softmax(score, axis=-1)
14         context_vec = att_weights * enc_output
15         context_vec = tf.reduce_sum(context_vec, axis=-1)
16         return context_vec, att_weights

```


In []:

```

1 class Onestepdecoder(tf.keras.Model):
2     '''THIS MODEL OUTPUTS THE RESULT OF DECODER FOR ONE TIME SETP GIVEN THE INPUT FOR F
3
4     def __init__(self, vocab_size,emb_dims, dec_units, input_len,att_units,batch_size):
5         super().__init__()
6         self.emb = layers.Embedding(vocab_size,emb_dims,input_length= input_len)
7         # ATTENTION LAYER
8         self.att = Attention(att_units)
9         # LSTM LAYER
10        self.lstm = layers.LSTM(dec_units,return_sequences=True,return_state=True)
11        # DENSE LAYER
12        self.dense = layers.Dense(vocab_size,activation="softmax")
13
14    def call(self, encoder_output , input , state_h,state_c):
15        emb = self.emb(input)
16
17        dec_output,dec_state_h,dec_state_c = self.lstm( emb , initial_state = [state_h,
18        context_vec,alphas = self.att(encoder_output,dec_state_h)
19        dense_input = tf.concat([tf.expand_dims(context_vec,1),dec_output],axis=-1)
20        fc = self.dense(dense_input)
21        return fc , dec_state_h , dec_state_c , alphas

```

###Decoder Class

In []:

```

1 class Decoder(tf.keras.Model):
2     '''THIS MODEL PERFORMS THE WHOLE DECODER OPERATION FOR THE COMPLETE SENTENCE'''
3     def __init__(self, vocab_size,emb_dims, dec_units, input_len,att_units,batch_size):
4         super().__init__()
5
6         self.input_len = input_len
7         self.onestepdecoder = Onestepdecoder(vocab_size,emb_dims, dec_units, input_len,
8
9     def call(self,dec_input,enc_output,state_h,state_c):
10        current_state_h = state_h
11        current_state_c = state_c
12        pred = []
13        alpha_values = []
14        for i in range(self.input_len):
15            current_vec = dec_input[:,i]
16            current_vec = tf.expand_dims(current_vec,axis=-1)
17            dec_output,dec_state_h,dec_state_c,alphas = self.onestepdecoder(enc_output
18            current_state_h = dec_state_h
19            current_state_c = dec_state_c
20            pred.append(dec_output)
21            alpha_values.append(alphas)
22        output = tf.concat(pred,axis=1)
23        alpha_values = tf.concat(alpha_values,axis = -1)
24        return output , alpha_values

```


In [50]:

```
1 %load_ext tensorboard
2 %tensorboard --logdir /content/drive/MyDrive/ColabNotebooks/cs2/model_save/attention_ge
```

The tensorboard extension is already loaded. To reload it, use:

```
%reload_ext tensorboard
```

<IPython.core.display.Javascript object>

In [56]:

```
1 model.build([(32,35),(32,35)])
2 model.summary()
```

Model: "encoder_decoder"

Layer (type)	Output Shape	Param #
=====		
encoder (Encoder)	multiple	10583768
decoder (Decoder)	multiple	22835938

```
=====
Total params: 33,419,706
Trainable params: 33,419,706
Non-trainable params: 0
=====
```

In []:

```
1 model.load_weights("/content/drive/MyDrive/ColabNotebooks/cs2/model_save/attention_gerr
```

This function is used in inference time which given any sentence in italian outputs the english sentence and alpha values

In []:

```

1 def predict(ita_text,model):
2
3
4     seq = tk_inp.texts_to_sequences([ita_text])
5     seq = pad_sequences(seq,maxlen = 20 , padding="post")
6     state = model.layers[0].initialize(1)
7     enc_output,state_h,state_c= model.layers[0](seq,state)
8     pred = []
9     input_state_h = state_h
10    input_state_c = state_c
11    current_vec = tf.ones((1,1))
12    alpha_values = []
13
14    for i in range(20):
15        fc , dec_state_h ,dec_state_c, alphas = model.layers[1].layers[0](enc_output ,
16            alpha_values.append(alphas)
17            current_vec = np.argmax(fc , axis = -1)
18            input_state_h = dec_state_h
19            input_state_c = dec_state_c
20            pred.append(tk_out.index_word[current_vec[0][0]])
21            if tk_out.index_word[current_vec[0][0]]=="<end>":
22                break
23    pred_sent = " ".join(pred)
24    alpha_values = tf.squeeze(tf.concat(alpha_values,axis=-1),axis=0)
25    return pred_sent , alpha_values

```

Calculating BLUE score

In []:

```

1 BLEU = []
2 np.random.seed(1)
3 test_data = df_val.loc[np.random.choice(df_val.index,size = 2000,replace=False)]
4 for ind,i in tqdm(test_data.iterrows(),position=0):
5     try:
6         pred = predict(str(i.enc_input),model)[0].split()
7         act = [str(i.dec_output).split()]
8         b = bleu.sentence_bleu(act,pred)
9         BLEU.append(b)
10    except:
11        continue
12 print("BLEU = ", np.mean(BLEU))

```

##Testing the model

In [51]:

```
1 print("INPUT SENTENCE ==> ", "it is not of only your business")
2 print("PREDICTED SENTENCE ==> ", predict("it is not of only your business", model)[0])
3 print("ACTUAL SENTENCE ==> ", "It is none of your business.")
```

INPUT SENTENCE ==> it is not of only your business
PREDICTED SENTENCE ==> it is not of only your business <end>
ACTUAL SENTENCE ==> It is none of your business.

In [53]:

```
1 print("INPUT SENTENCE ==> ", "She see Tom is caught by policeman in park at last night")
2 print("PREDICTED SENTENCE ==> ", predict("She see Tom is caught by policeman in park at last night", model)[0])
3 print("ACTUAL SENTENCE ==> ", "She saw Tom caught by a policeman in the park last night")
```

INPUT SENTENCE ==> She see Tom is caught by policeman in park at last night.
PREDICTED SENTENCE ==> she saw tom is caught by a policeman in the park at last night <end>
ACTUAL SENTENCE ==> She saw Tom caught by a policeman in the park last night.

Conclusion: If this model will train for a large number epoches and if proper dataset can be found this model will work better.

In []:

```
1
```