Imports

In [1]:

```
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
In [2]:
import pickle
train df = pickle.load(open('/content/drive/MyDrive/Copy of train df prep.pkl','rb'))
In [3]:
!pip install transformers
Collecting transformers
  Downloading transformers-4.15.0-py3-none-any.whl (3.4 MB)
                                      | 3.4 MB 4.2 MB/s
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from t
ransformers) (2.23.0)
Collecting sacremoses
  Downloading sacremoses-0.0.46-py3-none-any.whl (895 kB)
                                      | 895 \text{ kB } 55.2 \text{ MB/s}
Collecting pyyaml>=5.1
  Downloading PyYAML-6.0-cp37-cp37m-manylinux_2_5_x86_64.manylinux1_x86_64.manylinux_2_12
_{x86\_64.manylinux2010\_x86\_64.whl} (596 kB)
                                      | 596 kB 60.2 MB/s
Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packag
es (from transformers) (4.8.2)
Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from t
ransformers) (3.4.0)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.7/dist-packages
(from transformers) (21.3)
Collecting tokenizers<0.11,>=0.10.1
  Downloading tokenizers-0.10.3-cp37-cp37m-manylinux 2 5 x86 64.manylinux1 x86 64.manylin
ux_2_12_x86_64.manylinux2010_x86_64.whl (3.3 MB)
                                      | 3.3 \text{ MB } 48.7 \text{ MB/s}
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.7/dist-packages (fro
m transformers) (1.19.5)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.7/dist-packages (from
transformers) (4.62.3)
Collecting huggingface-hub<1.0,>=0.1.0
  Downloading huggingface_hub-0.2.1-py3-none-any.whl (61 kB)
                                     | 61 kB 411 kB/s
Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.7/dist-package
s (from transformers) (2019.12.20)
Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.7/dis
t-packages (from huggingface-hub<1.0,>=0.1.0->transformers) (3.10.0.2)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /usr/local/lib/python3.7/dist-
packages (from packaging>=20.0->transformers) (3.0.6)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from
importlib-metadata->transformers) (3.6.0)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-package
s (from requests->transformers) (3.0.4)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packag
es (from requests->transformers) (2021.10.8)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (fr
om requests->transformers) (2.10)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/
python3.7/dist-packages (from requests->transformers) (1.24.3)
Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from sacr
emoses->transformers) (7.1.2)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from sacrem
oses->transformers) (1.15.0)
Damiinamant aluandu astistist ishlik in /wan/lasal/lik/muthan0 7/dist mashanas /fwam asa
```

```
requirement aiready satisfied: jodito in /usr/focat/fid/python3.//dist-packages (from sac
remoses->transformers) (1.1.0)
Installing collected packages: pyyaml, tokenizers, sacremoses, huggingface-hub, transform
 Attempting uninstall: pyyaml
    Found existing installation: PyYAML 3.13
   Uninstalling PyYAML-3.13:
     Successfully uninstalled PyYAML-3.13
Successfully installed huggingface-hub-0.2.1 pyyaml-6.0 sacremoses-0.0.46 tokenizers-0.10
.3 transformers-4.15.0
In [ ]:
from transformers import BertTokenizer,DistilBertTokenizer
tokenizer2 = DistilBertTokenizer.from pretrained('distilbert-base-uncased',add prefix spa
                  # tokenizer for distill bert
ce=True)
In [4]:
from transformers import BertTokenizer, DistilBertTokenizer
```

```
# tokenizer for bert
tokenizer = BertTokenizer.from pretrained('bert-base-uncased',add prefix space=True)
```

In [5]:

from transformers import TFBertForQuestionAnswering, TFDistilBertForQuestionAnswering dbert= TFDistilBertForQuestionAnswering.from pretrained('distilbert-base-uncased') bert= TFBertForQuestionAnswering.from pretrained('bert-base-uncased')

All model checkpoint layers were used when initializing TFBertForQuestionAnswering.

Some layers of TFBertForQuestionAnswering were not initialized from the model checkpoint at bert-base-uncased and are newly initialized: ['qa_outputs'] You should probably TRAIN this model on a down-stream task to be able to use it for predi ctions and inference.

In [6]:

```
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Softmax, Dense, Activation, Dropout
from sklearn.model selection import train test split
import numpy as np
import pandas as pd
from tqdm import tqdm
import tensorflow as tf
from tensorflow.keras.callbacks import ModelCheckpoint
from tensorflow.keras.callbacks import CSVLogger, EarlyStopping, ModelCheckpoint, LearningR
ateScheduler
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import Sequential, Model
from tensorflow.keras.layers import Activation, Dropout, Flatten, Dense, Conv2D, MaxPool
ing2D, Embedding, LSTM, Embedding, Input, Softmax, Dense, Activation, Dropout
# from fuzzywuzzy import fuzz
from sklearn.preprocessing import LabelEncoder,OneHotEncoder
import time
from time import time
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad sequences
import matplotlib.pyplot as plt
import seaborn as sns
```

In [7]:

```
x_train, x_val , y_train , y_val = train_test_split(train_df[['text','sentiment']],train
df['selected text'], test size=0.2, random state=42)
```

```
x_train.shape, x_val.shape , y_train.shape , y_val.shape
Out[7]:
    ((21976, 2), (5495, 2), (21976,), (5495,))

metric

In []:
    def jaccard(str1,str2):
        a=set(str1.lower().split())
        b=set(str2.lower().split())
        c=a.intersection(b)
    return float(len(c)) / (len(a) + len(b) - len(c))
```

Models_comparision

pre_processing

text to input_ids

```
In [9]:
```

```
def train prep(MAX LEN, tokenizer):
                                                                            # train input
_ids and attten_mask start and end tokens indexes
 count = x train.shape[0]
 input ids = np.zeros((count, MAX LEN), dtype='int32')
 attention mask = np.zeros((count, MAX LEN), dtype='int32')
 start tokens = np.zeros((count, MAX LEN), dtype='int32')
 end_tokens = np.zeros((count,MAX_LEN),dtype='int32')
 toks all = []
 count=0
 for i, each in tqdm(enumerate(x train.values)):
   val = tokenizer.encode plus(each[0],each[1],add special tokens=True,max length=MAX L
EN, return attention mask=True, pad to max length=True, return tensors='tf', verbose=False)
   input ids[i] = val['input ids']
   attention mask[i] = val['attention mask']
    text1 = ""+" ".join(each[0].split())
    text2 = " ".join(y_train.values[i].split())
    idx = text1.find(text2)
    chars = np.zeros((len(text1)))
   chars[idx:idx+len(text2)]=1
   if text1[idx-1] == ' ':
      chars[idx-1] = 1
   enc = tokenizer.encode(text1)
   offsets = []; idx=0
    for t in enc:
     w=tokenizer.decode([t])
     offsets.append((idx,idx+len(w)))
     idx += len(w)
    toks = []
    for c,(a,b) in enumerate(offsets):
      sm = np.sum(chars[a:b])
      if sm>0:
       toks.append(c)
    toks all.append(toks)
    if len(toks)>0:
      count+=1
      start tokens[i,(toks[0])+1] = 1
      end tokens[i, (toks[-1])+1] = 1
```

```
return toks all,input ids,attention mask,start tokens,end tokens
```

In [11]:

```
toks_all,input_ids,attention_mask,start_tokens,end_tokens = train_prep(MAX_LEN=92,tokeni
zer=tokenizer) # train ids
```

In []:

```
def valid prep(MAX LEN, tokenizer):
  count = y val.shape[0]
 input ids val = np.zeros((count, MAX LEN), dtype='int32')
  attention_mask_val = np.zeros((count, MAX_LEN), dtype='int32')
 start tokens val = np.zeros((count, MAX LEN), dtype='int32')
  end_tokens_val = np.zeros((count,MAX_LEN),dtype='int32')
  for i, each in tqdm(enumerate(x val.values)):
   val = tokenizer.encode plus(each[0],each[1],add special tokens=True,max length=MAX L
EN, return attention mask=True, pad to max length=True, return tensors='tf', verbose=False)
   input ids val[i] = val['input ids']
   attention mask val[i] = val['attention mask']
   text1 = " "+" ".join(each[0].split())
   text2 = " ".join(y_val.values[i].split())
    #finding the start index
   idx = text1.find(text2)
   chars = np.zeros((len(text1)))
   chars[idx:idx+len(text2)]=1
   if text1[idx-1]==' ':
     chars[idx-1] = 1
   enc = tokenizer.encode(text1)
   offsets = []; idx=0
   for t in enc:
     w=tokenizer.decode([t])
      offsets.append((idx,idx+len(w)))
     idx += len(w)
    toks = []
    for c, (a,b) in enumerate(offsets):
     sm = np.sum(chars[a:b])
      if sm>0:
       toks.append(c)
    toks all.append(toks)
    if len(toks)>0:
     count+=1
      start tokens val[i, (toks[0])+1] = 1
      end_tokens_val[i,(toks[-1])+1] = 1
  return input ids val, attention mask val, start tokens val, end tokens val
```

In []:

```
input_ids_val,attention_mask_val,start_tokens_val,end_tokens_val = valid_prep(MAX_LEN=92,
tokenizer=tokenizer) # valid ids
5495it [00:09, 586.80it/s]
```

In [15]:

```
input_data = (input_ids,attention_mask)
output_data = (start_tokens,end_tokens)
val = (input_ids_val,attention_mask_val)
output_val = (start_tokens_val,end_tokens_val)
val_data = (val,output_val)

train_dataset = tf.data.Dataset.from_tensor_slices((input_data, output_data)).shuffle(bu
ffer_size=1024).batch(32)
```

```
val_dataset = tf.data.Dataset.from_tensor_slices(val_data).batch(32)
```

DBERT_QA+dense

In []:

```
MAX LEN=128
# reload
def create_model():
 input1 = Input(shape=(MAX_LEN,),name='input_id',dtype=tf.int32)
  input2 = Input(shape=(MAX_LEN,),name='attention_mask',dtype=tf.int32)
  start scores, end scores = dbert(input1, attention mask = input2).values()
  dense1 = Dense(units=MAX LEN, activation='relu', name='dense1', kernel regularizer = tf.k
eras.regularizers.L2(12=0.00001))(start scores)
  softmax1 = Activation('softmax')(densel)
  dense2 = Dense(units=MAX_LEN,activation='relu',name='dense2',kernel_regularizer = tf.k
eras.regularizers.L2(12=0.00001))(end scores)
 softmax2 = Activation('softmax')(dense2)
 model = Model(inputs=[input1,input2],outputs=[softmax1,softmax2])
 opt = tf.keras.optimizers.Adam(learning rate=0.0001, epsilon=1e-08, clipnorm=1.0)
 model.compile(optimizer=opt,loss='categorical crossentropy')
 return model
model = create model()
model.load weights('/content/drive/MyDrive/Copy of DBertQA model.hdf5')
```

In []:

activation 7 (Activation)

model.summary()			
Model: "model_3"			
Layer (type)	Output Shape		
======================================		0	[]
attention_mask (InputLayer)	[(None, 128)]	0	[]
<pre>tf_distil_bert_for_question_an swering (TFDistilBertForQuesti] onAnswering)</pre>		66364418	<pre>['input_id[0][0]', 'attention_mask[0][0]'</pre>
densel (Dense) estion_ans	(None, 128)	16512	<pre>['tf_distil_bert_for_qu wering[3][1]']</pre>
<pre>dense2 (Dense) estion_ans</pre>	(None, 128)	16512	<pre>['tf_distil_bert_for_qu wering[3][0]']</pre>
activation_6 (Activation)	(None, 128)	0	['dense1[0][0]']

/NTODO 1001

[14076026011

```
activation / (Activation) (None, 120) v [ densez[v][v] ]
```

Total params: 66,397,442
Trainable params: 66,397,442
Non-trainable params: 0

predictions

In []:

```
val = (input_ids_val,attention_mask_val)
start val , end val = model.predict(val)
                                                       # validation predictions
pred values val=[]
for i in tqdm(range(start val.shape[0])):
 a = np.argmax(start val[i])
 b = np.argmax(end val[i])
 text1 = " "+" ".join(x_val['text'].values[i].split())
                                                                     # pred answer
 enc = tokenizer2.encode(text1)
 val = tokenizer2.decode(enc[a:b+1])
 pred values val.append(val)
for i in range(len(pred values val)):
 pred_values_val[i] = pred_values_val[i].replace('[SEP]','') # removing [sep] token
scores val=[]
x val['pred text'] = pred values val
x_val['selected_text'] = y_val.values
scores=[]
for i in tqdm(range(x_val.shape[0])):
 scores_val.append(jaccard(x_val['pred_text'].values[i],x_val['selected_text'].values[i
             # jaccard score calculations
]))
x val['jaccard']=scores val
```

In []:

```
x val.head(20)
```

jaccard	selected_text	pred_text	sentiment	text	
1.000000	this is my update	this is my update	neutral	this is my update	7917
1.000000	whaaat i still have next week	whaaat i still have next week	neutral	whaaat i still have next week	15845
0.600000	happy birthday little sister of mine	happy birthday little sister of mine also good	positive	happy birthday little sister of mine also go	21278
1.000000	do some research for my article	do some research for my article	neutral	do some research for my article	22338
1.000000	are you okay	are you okay	neutral	are you okay	22474
0.043478	fail	turned my alarm off this morning because i tho	negative	turned my alarm off this morning because i tho	21560
1.000000	she gives good advice	she gives good advice	positive	elaines my online mommy too she gives good a	8870
0.333333	good	good abuse ho	positive	good ABUSE homie hahahahaha thats what im tal	22030
0.590909	the annual problem of r generation is finding	there are days of summervac school comes along	neutral	there are days of summervac school comes alo	6693
1.000000	i have tea have just found a picture of the b	i have tea have just found a picture of the bi	neutral	i have tea have just found a picture of the b	15718
1.000000	craving coffee	craving coffee	neutral	craving coffee	15279

20672	time for tv in bedthen spending all day catchi	s ertjale ra	online summ eprela<u>s</u>teat	i hate online sur seterteld<u>s</u>text	0 j600000
17239	well it almost was a good day guess i just ret	positive	good day guess	good day	0.666667
26658	the ultimate shirt folding tool i saw using	positive	the ultimate shirt folding tool i saw using th	ultimate	0.066667
6199	no months	neutral	no months	no months	1.000000
10794	the least i can do for you is retweet it when	neutral	the least i can do for you is retweet it when	the least i can do for you is retweet it when	1.000000
2899	oh snap kinda nuts right now ive told at lea	positive	oh snap kinda nuts right now ive told at least	thanks	0.083333
12789	says gud eve guys lets play poker yeah cant re	positive	yeah cant read	says gud eve guys	0.000000
5036	awesome im glad you like it fyi platinum no	positive	awesome im	glad	0.000000
2938	im celebrating my mother and also celebrating	positive	im celebrating my	celebrating	0.333333

In []:

```
print('Mean training Jaccard score:',np.mean(x_val['jaccard']))
print("="*150)
print('Mean jaccard score for positive sentiment tweets:',np.mean(x_val[x_val['sentiment']=='positive']['jaccard']))
print("="*150)
print('Mean jaccard score for negative sentiment tweets',np.mean(x_val[x_val['sentiment']=='negative']['jaccard']))
print("="*150)
print('Mean jaccard score for neutral sentiment tweets',np.mean(x_val[x_val['sentiment']=='neutral']['jaccard']))
```

Mean training Jaccard score: 0.6101828553749853

Mean jaccard score for positive sentiment tweets: 0.3745756631390002

Mean jaccard score for negative sentiment tweets 0.35826182939871276

Mean jaccard score for neutral sentiment tweets 0.9672193330825384

DBERT_QA+conv1d

In []:

```
MAX LEN = 92
def DBERT CNN():
  input1 = Input(shape=(MAX LEN,),name='input_id',dtype=tf.int32)
  input2 = Input(shape=(MAX LEN,),name='attention mask',dtype=tf.int32)
  start scores, end scores = dbert(input1, attention mask = input2).values()
  drop1 = Dropout(0.1)(start scores)
  drop1 = tf.expand_dims(drop1,axis=-1)
  layer1 = tf.keras.layers.Conv1D(1,1)(drop1)
# convld instead dense
  layer1= Flatten()(layer1)
  softmax1 = Activation('softmax')(layer1)
  drop2 = Dropout(0.1) (end scores)
  drop2 = tf.expand_dims(drop2,axis=-1)
  layer2 = tf.keras.layers.Conv1D(1,1)(drop2)
conv1d
  layer2 = Flatten()(layer2)
  softmax2 = Activation('softmax')(layer2)
```

```
model = Model(inputs=[input1,input2],outputs=[softmax1,softmax2])

opt = tf.keras.optimizers.Adam(learning_rate=3e-5, epsilon=1e-08, clipnorm=1.0)
model.compile(optimizer=opt,loss='categorical_crossentropy')

return model

model = DBERT_CNN()

model.load_weights('/content/drive/MyDrive/DBert_CNN_QA_model.hdf5')
```

predictions

In []:

```
val = (input ids val, attention mask val)
start val , end val = model.predict(val)
                                                       # validation predictions
pred values val=[]
for i in tqdm(range(start val.shape[0])):
 a = np.argmax(start val[i])
 b = np.argmax(end_val[i])
 text1 = " "+" ".join(x_val['text'].values[i].split())
                                                                    # pred answer
 enc = tokenizer2.encode(text1)
 val = tokenizer2.decode(enc[a:b+1])
 pred values val.append(val)
for i in range(len(pred_values_val)):
 pred values val[i] = pred values val[i].replace('[SEP]','') # removing [sep] token
scores val=[]
x val['pred text'] = pred values val
x val['selected_text'] = y_val.values
scores=[]
for i in tqdm(range(x val.shape[0])):
 scores val.append(jaccard(x val['pred text'].values[i],x val['selected text'].values[i
             # jaccard score calculations
x val['jaccard']=scores val
```

In []:

```
model.summary()
```

Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to		
 ======= input_id (InputLayer)	[(None, 92)]	0	[]		
attention_mask (InputLayer)	[(None, 92)]	0	[]		
<pre>tf_distil_bert_for_question_an swering (TFDistilBertForQuesti)</pre>			['input_id[0][0]',		
onAnswering)	ne, start_logits=(N		decention_mask[0][0]]		
	one, 92),				
	end_logits=(None,				
	92),				
	hidden_states=None				

, attentions=None)

<pre>dropout_24 (Dropout) estion_ans</pre>	(None, 92)	0	<pre>['tf_distil_bert_for_qu wering[2][1]']</pre>
<pre>dropout_25 (Dropout) estion_ans</pre>	(None, 92)	0	<pre>['tf_distil_bert_for_qu wering[2][0]']</pre>
tf.expand_dims_4 (TFOpLambda)	(None, 92, 1)	0	['dropout_24[0][0]']
tf.expand_dims_5 (TFOpLambda)	(None, 92, 1)	0	['dropout_25[0][0]']
<pre>conv1d_4 (Conv1D)]']</pre>	(None, 92, 1)	2	['tf.expand_dims_4[0][0
conv1d_5 (Conv1D)	(None, 92, 1)	2	['tf.expand_dims_5[0][0
flatten_4 (Flatten)	(None, 92)	0	['conv1d_4[0][0]']
flatten_5 (Flatten)	(None, 92)	0	['conv1d_5[0][0]']
activation_4 (Activation)	(None, 92)	0	['flatten_4[0][0]']
activation_5 (Activation)	(None, 92)	0	['flatten_5[0][0]']

=======

Total params: 66,364,422 Trainable params: 66,364,422 Non-trainable params: 0

In []:

 $x_val.head(20)$

	text	sentiment	pred_text	selected_text	jaccard
7917	this is my update	neutral	this is my update	this is my update	1.000000
15845	whaaat i still have next week	neutral	whaaat i still have next week	whaaat i still have next week	1.000000
21278	happy birthday little sister of mine also go	positive	happy birthday little	happy birthday little sister of mine	0.500000
22338	do some research for my article	neutral	do some research for my article	do some research for my article	1.000000

22474	are you dead	sentiment	are pred<u>o</u>leas	sale glad_deag	1 jaccacd	
21560	turned my alarm off this morning because i tho	negative	turned my alarm off this morning because i tho	fail	0.043478	
8870	elaines my online mommy too she gives good a	positive	good advice	she gives good advice	0.500000	
22030	good ABUSE homie hahahahaha thats what im tal	positive	good	good	1.000000	
6693	there are days of summervac school comes alo	neutral	there are days of summervac school comes along	the annual problem of r generation is finding	0.590909	
15718	i have tea have just found a picture of the b	neutral	i have tea have just found a picture of the bi	i have tea have just found a picture of the b	1.000000	
15279	craving coffee	neutral	craving coffee	craving coffee	1.000000	
20672	time for tv in bedthen spending all day catchi	negative	online summer classes	i hate online summer classes	0.600000	
17239	well it almost was a good day guess i just ret	positive	day guess	good day	0.333333	
26658	the ultimate shirt folding tool i saw using	positive	the ultimate shirt folding tool i saw using th	ultimate	0.062500	
6199	no months	neutral	no months	no months	1.000000	
10794	the least i can do for you is retweet it when	neutral	the least i can do for you is retweet it when	the least i can do for you is retweet it when	1.000000	
2899	oh snap kinda nuts right now ive told at lea	positive	kinda nuts right now ive told at least thanks	thanks	0.100000	
12789	says gud eve guys lets play poker yeah cant re	positive	##t read my poker face	says gud eve guys	0.000000	
5036	awesome im glad you like it fyi platinum no	positive	awesome im	glad	0.000000	
2938	im celebrating my mother and also celebrating	positive	im celebrating my	celebrating	0.333333	
In []	:					
<pre>print('Mean training Jaccard score:',np.mean(x_val['jaccard'])) print("="*150) print('Mean jaccard score for positive sentiment tweets:',np.mean(x_val[x_val['sentiment']="positive']['jaccard'])) print("="*150) print('Mean jaccard score for negative sentiment tweets',np.mean(x_val[x_val['sentiment']="negative']['jaccard'])) print("="*150) print('Mean jaccard score for neutral sentiment tweets',np.mean(x_val[x_val['sentiment']="neutral']['jaccard']))</pre>						
Mean training Jaccard score: 0.6107009628532492						
Mean jaccard score for positive sentiment tweets: 0.36968854032180715						
Mean jaccard score for negative sentiment tweets 0.35657944122532753						
	accard score for neutral			42162		
In []	:					

BERT_QA+conv1d

```
def BERT QA():
 MAX LEN = 92
  input1 = Input(shape=(MAX_LEN,),name='input_id',dtype=tf.int32)
 input2 = Input(shape=(MAX_LEN,),name='attention_mask',dtype=tf.int32)
 start scores, end scores = bert(input1, attention_mask = input2).values()
  drop1 = Dropout(0.5)(start scores)
  drop1 = tf.expand dims(drop1,axis=-1)
 layer1 = tf.keras.layers.Conv1D(1,1)(drop1)
# convld instead dense
 layer1= Flatten()(layer1)
  softmax1 = Activation('softmax')(layer1)
  drop2 = Dropout(0.5)(end_scores)
  drop2 = tf.expand_dims(drop2,axis=-1)
  layer2 = tf.keras.layers.Conv1D(1,1)(drop2)
conv1d
 layer2 = Flatten()(layer2)
  softmax2 = Activation('softmax')(layer2)
 model = Model(inputs=[input1,input2],outputs=[softmax1,softmax2])
 opt = tf.keras.optimizers.Adam(learning rate=2e-5, epsilon=1e-08, clipnorm=1.0)
 model.compile(optimizer=opt,loss='categorical crossentropy')
 return model
model = BERT QA()
model.load weights('/content/drive/MyDrive/Copy of Copy of Bert CNN QA model.hdf5')
```

In []:

```
model.summary()
```

Model: "model_5"						
Layer (type)	Output Shape	Param #	Connected to			
======= input_id (InputLayer)	[(None, 92)]	0	[]			
attention_mask (InputLayer)	[(None, 92)]	0	[]			
tf_bert_for_question_answering	TFQuestionAnswering	108893186	['input_id[0][0]',			
(TFBertForQuestionAnswering)	ModelOutput(loss=No		'attention_mask[0][0]']			
	ne, start_logits=(N					
	one, 92),					
	end_logits=(None,					
	92),					
	hidden_states=None					
	, attentions=None)					
<pre>dropout_85 (Dropout) answering[</pre>	(None, 92)	0	['tf_bert_for_question_			

1][1]']

```
dropout 86 (Dropout)
                  (None, 92)
                                                     ['tf bert for question
answering[
                                                     1][0][]
tf.expand dims 8 (TFOpLambda) (None, 92, 1)
                                          0
                                                     ['dropout 85[0][0]']
                          (None, 92, 1)
tf.expand_dims_9 (TFOpLambda)
                                           0
                                                     ['dropout_86[0][0]']
                                           2
conv1d 8 (Conv1D)
                          (None, 92, 1)
                                                     ['tf.expand dims 8[0][0
]']
                                                     ['tf.expand dims_9[0][0
conv1d 9 (Conv1D)
                          (None, 92, 1)
flatten 8 (Flatten)
                          (None, 92)
                                           0
                                                     ['conv1d 8[0][0]']
flatten 9 (Flatten)
                         (None, 92)
                                            0
                                                     ['conv1d 9[0][0]']
activation 10 (Activation) (None, 92)
                                          0
                                                     ['flatten 8[0][0]']
activation 11 (Activation) (None, 92)
                                   0
                                                     ['flatten 9[0][0]']
_______
```

Non-trainable params: 0

Total params: 108,893,190
Trainable params: 108,893,190

train performance

In []:

```
input data = (input ids,attention mask)
start , end = model.predict(input_data)
pred values=[]
for i in tqdm(range(start.shape[0])):
 a = np.argmax(start[i])
 b = np.argmax(end[i])
 text1 = " "+" ".join(x_train['text'].values[i].split())
 enc = tokenizer.encode(text1)
 val = tokenizer.decode(enc[a:b+1])
 pred_values.append(val)
for i in range(len(pred_values)):
 pred values[i] = pred values[i].replace('[SEP]','') # removing [sep] token
x train['pred text'] = pred values
x train['selected text'] = y train.values
scores=[]
for i in tqdm(range(x train.shape[0])):
 scores.append(jaccard(x_train['pred_text'].values[i],x_train['selected_text'].values[i
]))
```

```
x_train['jaccard']=scores
100%| 21976/21976 [00:18<00:00, 1181.78it/s]
```

In []:

 $x_{train.head(20)}$

Out[]:

	text	sentiment	pred_text	selected_text	jaccard
22723	awe thanks	positive	awe thanks	awe thanks	1.000000
1231	on my way to dazzle bar	neutral	on my way to dazzle bar	on my way to dazzle bar	1.000000
531	terrible haha that was new the fact that he c	neutral	terrible haha that was new the fact that he co	he couldnt do it should have cost his part	0.428571
22325	fun day with boo short but fun	positive	fun day	fun day	1.000000
18214	of course im almost there and theres traffic	neutral	of course im almost there and theres traffic	of course im almost there and theres traffic	1.000000
15839	im nauseous i need a yogurt or something	negative	im nauseous	nauseous	0.500000
17356	omg my head still hurts i need to get my comic	negative	hurts i	hurts	0.500000
12502	my body is aching i can barely movebut its wor	negative	my body is aching i can barely movebu	barely	0.125000
25623	oh no my fan broke nooooooooooo great now i	negative	##000000000	noooooooooo	0.000000
12306	are you tinkn the people on the street in ny	neutral	are you tinkn the people on the street in ny w	are you tinkn the people on the street in ny w	1.000000
2863	u no that little prob with ur twitter that ha	positive	this will help u fix it	mayb this will help u fix it	0.857143
7457	to my pretty lady happy mothers day shes the	positive	happy mothers	happy	0.500000
21656	its a an mk mentalityi think	neutral	its a an mk mentalityi think	its a an mk mentalityi think	1.000000
15772	going to shibuya meet my mom get my contact le	neutral	going to shibuya meet my mom get my contact le	going to shibuya meet my mom get my contact le	1.000000
14740	thanks i love the word crikey its like my sayin	positive	thanks i	thanks	0.500000
7376	getting bored of walking up and down the stairs	negative	getting bored of	getting bored of walking up and down the stairs	0.333333
18461	feeling pretty tired and lonely	negative	feeling pretty tired and lonely	feeling pretty tired and lonely	1.000000
5976	ohhh how lovely im glad that you had a grea	positive	ohhh how lovely im glad that you had a great t	ohhh how lovely im glad that you had a great t	1.000000
12647	i made roomi dont have a bed from how much ro	negative	i made roomi dont have a bed from how much roo	outta space	0.095238
15760	ABUSEpoints at facebook messageABUSE	neutral	abusepoints at facebook messageabuse	ABUSEpoints at facebook messageABUSE	1.000000

In []:

```
print('Mean training Jaccard score:',np.mean(x_train['jaccard']))
print("="*150)
print('Mean jaccard score for positive sentiment tweets:',np.mean(x_train[x_train['sentiment']=='positive']['jaccard']))
print("="*150)
print('Mean jaccard score for negative sentiment tweets',np.mean(x_train[x_train['sentiment']=='negative']['jaccard']))
print("="*150)
print('Mean jaccard score for neutral sentiment tweets',np.mean(x_train[x_train['sentiment']=='neutral']['jaccard']))
```

Validation_performance

In []:

```
val = (input ids val, attention mask val)
start_val , end_val = model.predict(val)
                                                        # validation predictions
pred values val=[]
for i in tqdm(range(start_val.shape[0])):
 a = np.argmax(start val[i])
 b = np.argmax(end val[i])
 text1 = " "+" ".join(x_val['text'].values[i].split())
                                                                     # pred answer
 enc = tokenizer2.encode(text1)
 val = tokenizer2.decode(enc[a:b+1])
 pred values val.append(val)
for i in range(len(pred_values_val)):
 pred_values_val[i] = pred_values_val[i].replace('[SEP]','') # removing [sep] token
scores val=[]
x val['pred text'] = pred values val
x_val['selected_text'] = y_val.values
scores=[]
for i in tqdm(range(x val.shape[0])):
 scores_val.append(jaccard(x_val['pred_text'].values[i],x val['selected text'].values[i
              # jaccard score calculations
]))
x val['jaccard']=scores val
```

In []:

```
x_val.head(20)
```

	text	sentiment	pred_text	selected_text	jaccard
7917	this is my update	neutral	this is my update	this is my update	1.000000
15845	whaaat i still have next week	neutral	whaaat i still have next week	whaaat i still have next week	1.000000
21278	happy birthday little sister of mine also go	positive	happy birthday	happy birthday little sister of mine	0.333333
22338	do some research for my article	neutral	do some research for my article	do some research for my article	1.000000
22474	are you okay	neutral	are you okay	are you okay	1.000000
21560	turned my alarm off this morning because i tho	negative		fail	0.000000
8870	elaines my online mommy too she gives good a	positive	good advice	she gives good advice	0.500000
22030	good ABUSE homie hahahahaha thats what im tal	positive	good	good	1.000000
6693	there are days of summervac school comes alo	neutral	there are days of summervac school comes along	the annual problem of r generation is finding	0.590909
15718	i have tea have just found a picture of the b	neutral	i have tea have just found a picture of the bi	i have tea have just found a picture of the b	1.000000
15279	craving coffee	neutral	craving coffee	craving coffee	1.000000
20672	time for tv in bedthen spending all	nenative	online summer	i hate online summer classes	0 400000

	day catchi text	sentiment	pred_text	selected_text	jaccard
17239	well it almost was a good day guess i just ret	positive	day guess	good day	0.333333
26658	the ultimate shirt folding tool i saw using	positive	the ultimate shirt folding tool i saw using th	ultimate	0.062500
6199	no months	neutral	no months	no months	1.000000
10794	the least i can do for you is retweet it when	neutral	the least i can do for you is retweet it when	the least i can do for you is retweet it when	1.000000
2899	oh snap kinda nuts right now ive told at lea	positive	babes	thanks	0.000000
12789	says gud eve guys lets play poker yeah cant re	positive	cant read my poker face	says gud eve guys	0.000000
5036	awesome im glad you like it fyi platinum no	positive	awesome im	glad	0.000000
2938	im celebrating my mother and also celebrating	positive	im celebrating my mother and also celebrating	celebrating	0.083333

```
In [ ]:
```

```
print('Mean training Jaccard score:',np.mean(x val['jaccard']))
print("="*150)
print('Mean jaccard score for positive sentiment tweets:',np.mean(x val[x val['sentiment'
] == 'positive']['jaccard']))
print("="*150)
print('Mean jaccard score for negative sentiment tweets',np.mean(x val[x val['sentiment']
=='negative']['jaccard']))
print("="*150)
print('Mean jaccard score for neutral sentiment tweets',np.mean(x_val[x_val['sentiment']=
='neutral']['jaccard']))
Mean training Jaccard score: 0.6125855286682741
______
_____
Mean jaccard score for positive sentiment tweets: 0.3721725002502277
______
______
```

observations

- 1. bertQA performing slight better than distill bertQA
- 2. bertQA have train jaccard 0.69233 and validation jaccard 0.6125 . so there isn't much over fitting .

Mean jaccard score for negative sentiment tweets 0.36163770200012707

Mean jaccard score for neutral sentiment tweets 0.972672837052568

Quantization

```
In [ ]:
```

```
!pip install -q tensorflow
!pip install -q tensorflow-model-optimization
```

```
| 213 kB 5.2 MB/s
```

```
In [ ]:
```

```
import tensorflow as tf
import numpy as np
import tensorflow_model_optimization as tfmot
import tempfile
```

```
In [ ]:
```

```
MAX_LEN = 92
  input1 = Input(shape=(MAX LEN,),name='input id',dtype=tf.int32)
  input2 = Input(shape=(MAX LEN,),name='attention mask',dtype=tf.int32)
  start scores, end scores = bert(input1, attention mask = input2).values()
  drop1 = Dropout(0.5)(start scores)
  drop1 = tf.expand dims(drop1,axis=-1)
  layer1 = tf.keras.layers.Conv1D(1,1)(drop1)
# conv1d instead dense
 layer1= Flatten()(layer1)
  softmax1 = Activation('softmax')(layer1)
  drop2 = Dropout(0.5) (end scores)
  drop2 = tf.expand dims(drop2,axis=-1)
  layer2 = tf.keras.layers.Conv1D(1,1)(drop2)
conv1d
  layer2 = Flatten()(layer2)
  softmax2 = Activation('softmax')(layer2)
  model = Model(inputs=[input1,input2],outputs=[softmax1,softmax2])
  opt = tf.keras.optimizers.Adam(learning_rate=2e-5, epsilon=1e-08, clipnorm=1.0)
  model.compile(optimizer=opt,loss='categorical crossentropy')
  return model
model = BERT QA()
model.load weights('/content/drive/MyDrive/Copy of Copy of Bert CNN QA model.hdf5')
In [ ]:
In [ ]:
In [ ]:
# model = tf.keras.models.load model('model.h5')
converter = tf.lite.TFLiteConverter.from keras model(model)
converter.target_spec.supported_ops = [tf.lite.OpsSet.TFLITE_BUILTINS, tf.lite.OpsSet.SE
LECT TF OPS]
tflite model = converter.convert()
#saving converted model in "converted model.tflite" file
open("converted model.tflite", "wb").write(tflite model)
WARNING:absl:Found untraced functions such as embeddings layer call fn, embeddings layer
call_and_return_conditional_losses, encoder_layer_call_fn, encoder_layer_call_and_return_
conditional_losses, LayerNorm_layer_call_fn while saving (showing 5 of 1040). These funct
ions will not be directly callable after loading.
INFO:tensorflow:Assets written to: /tmp/tmpgw8ct29s/assets
INFO:tensorflow:Assets written to: /tmp/tmpqw8ct29s/assets
INFO:absl:Using new converter: If you encounter a problem please file a bug. You can opt-
out by setting experimental new converter=False
WARNING:absl:Buffer deduplication procedure will be skipped when flatbuffer library is no
t properly loaded
Out[]:
434567104
In [ ]:
# model = tf.keras.models.load model('model.h5')
converter = tf.lite.TFLiteConverter.from keras model(model)
converter.optimizations = [tf.lite.Optimize.DEFAULT]
converter.target spec.supported ops = [tf.lite.OpsSet.TFLITE BUILTINS, tf.lite.OpsSet.SE
LECT_TF_OPS]
tflite quant model = converter.convert()
```

def BERT QA():

```
open("converted_quant_model.tflite", "wb").write(tflite_quant_model)
WARNING:absl:Found untraced functions such as embeddings layer call fn, embeddings layer
call and return conditional losses, encoder layer call fn, encoder layer call and return
conditional_losses, LayerNorm_layer_call_fn while saving (showing 5 of 1040). These funct
ions will not be directly callable after loading.
INFO:tensorflow:Assets written to: /tmp/tmpiz72wso1/assets
INFO:tensorflow:Assets written to: /tmp/tmpiz72wso1/assets
INFO: absl: Using new converter: If you encounter a problem please file a bug. You can opt-
out by setting experimental new converter=False
WARNING:absl:Buffer deduplication procedure will be skipped when flatbuffer library is no
t properly loaded
Out[]:
109450336
In [ ]:
import os
print("Float model in Mb:", os.path.getsize('converted model.tflite') / float(2**20))
print("Quantized model in Mb:", os.path.getsize('converted quant model.tflite') / float(
print("Compression ratio:", os.path.getsize('converted model.tflite')/os.path.getsize('c
onverted quant model.tflite'))
Float model in Mb: 414.43548583984375
Quantized model in Mb: 104.38265991210938
Compression ratio: 3.970348007885602
In [ ]:
import tensorflow as tf
import pickle
# val = pickle.load(open('/content/drive/MyDrive/quantization/val.pkl','rb'))
# Load TFLite model and allocate tensors.
interpreter = tf.lite.Interpreter(model path="/content/drive/MyDrive/quantization/convert
ed quant model.tflite")
interpreter.allocate tensors()
# Get input and output tensors.
input details = interpreter.get input details()
output details = interpreter.get_output_details()
# Test model on some input data.
input shape = input details[0]['shape']
start output = np.zeros(shape=(len(val[0]),92))
end output = np.zeros(shape=(len(val[0]),92))
for i in range(len(val[0])):
  input data0 = val[0][i].reshape(input shape)
  input data1 = val[1][i].reshape(input shape)
  interpreter.set_tensor(input_details[1]['index'], input_data0)
  interpreter.set_tensor(input_details[0]['index'], input data1)
  interpreter.invoke()
  start output[i,:] = interpreter.get tensor(output details[0]['index'])
  end output[i,:] = interpreter.get tensor(output details[1]['index'])
In [ ]:
pred values val=[]
from tqdm import tqdm
for i in tqdm(range(start output.shape[0])):
  a = np.argmax(start output[i])
  b = np.argmax(end output[i])
  text1 = " "+" ".join(x val['text'].values[i].split())
                                                                      # pred answer
  enc = tokenizer.encode(text1)
  val = tokenizer.decode(enc[a:b+1])
  pred values val.append(val)
```

#saving converted model in "converted_quant_model.tflite" file

| 5495/5495 [00:03<00:00, 1584.55it/s]

100%|

```
In []:
for i in range(len(pred_values_val)):
   pred_values_val[i] = pred_values_val[i].replace('[SEP]','') # removing [sep] token
```

```
In [ ]:
```

In []:

 x_val

Out[]:

	text	sentiment	pred_text	selected_text	jaccard
7917	this is my update	neutral	this is my update	this is my update	1.000000
15845	whaaat i still have next week	neutral	whaaat i still have next week	whaaat i still have next week	1.000000
21278	happy birthday little sister of mine also go	positive	happy birthday	happy birthday little sister of mine	0.333333
22338	do some research for my article	neutral	do some research for my article	do some research for my article	1.000000
22474	are you okay	neutral	are you okay	are you okay	1.000000
				•••	
6816	has a gym day and is hoping to enjoy the last	positive	to enjoy the	enjoy	0.333333
8829	pappadeux has some yummy strawberry lemonade	positive	##ux has some yu	yummy strawberry lemonade	0.000000
24019	omg i saw ur update nd it said david archulet	positive		awsome	0.000000
14448	starting tm alex and i are doing a whole week	neutral	starting tm alex and i are doing a whole week	starting tm alex and i are doing a whole week	1.000000
12569	people r weird	negative	r weird	people r weird	0.666667

5495 rows × 5 columns

In []:

```
print('Mean training Jaccard score:',np.mean(x_val['jaccard']))
print("="*150)
print('Mean jaccard score for positive sentiment tweets:',np.mean(x_val[x_val['sentiment']=='positive']['jaccard']))
print("="*150)
print('Mean jaccard score for negative sentiment tweets',np.mean(x_val[x_val['sentiment']=='negative']['jaccard']))
print("="*150)
print('Mean jaccard score for neutral sentiment tweets',np.mean(x_val[x_val['sentiment']=='neutral']['jaccard']))
```

Mean training Jaccard score: 0.6195375047976922

Mean jaccard score for positive sentiment tweets: 0.3849089228986509

Mean jaccard score for neutral sentiment tweets 0.9735434417562145

post training analysis

```
In [ ]:
```

```
import pickle
pickle.dump(x_val,open('/content/drive/MyDrive/quantization/x_val.pkl','wb'))
```

In [12]:

```
import pickle
x_val = pickle.load(open('/content/drive/MyDrive/quantization/x_val.pkl','rb'))
```

In [13]:

```
x_val['Num_words_ST'] = x_val['selected_text'].apply(lambda x:len(str(x).split()))
# Number Of words in Selected Text
x_val['Num_word_text'] = x_val['text'].apply(lambda x:len(str(x).split()))
# Number Of words in main text
x_val['difference_in_words'] = x_val['Num_word_text'] - x_val['Num_words_ST']
# difference in number of words
x_val['Num_words_PT'] = x_val['pred_text'].apply(lambda x:len(str(x).split()))
# Number Of words in Predicted Text
x_val['difference_words_ST&PT'] = x_val['Num_words_ST'] - x_val['Num_words_PT']
# difference number of words btw selected text &pred_text
```

In []:

x_val

	text	sentiment	pred_text	selected_text	jaccard	Num_words_PT	Num_words_ST	Num_word_text	difference
7917	this is my update	neutral	this is my update	this is my update	1.000000	4	4	4	
15845	whaaat i still have next week	neutral	whaaat i still have next week	whaaat i still have next week	1.000000	6	6	6	
21278	happy birthday little sister of mine also go	positive	happy birthday	happy birthday little sister of mine	0.333333	2	6	10	
22338	do some research for my article	neutral	do some research for my article	do some research for my article	1.000000	6	6	6	
22474	are you okay	neutral	are you okay	are you okay	1.000000	3	3	3	
6816	has a gym day and is hoping to enjoy the last	positive	to enjoy the	enjoy	0.333333	3	1	15	
8829	pappadeux has some yummy strawberry	positive	##ux has some yu	yummy strawberry lemonade	0.000000	4	3	6	

	iemonade text	sentiment	pred_text	selected_text	jaccard	Num_words_PT	Num_words_ST	Num_word_text	difference_
24019	omg i saw ur update nd it said david archulet	positive		awsome	0.000000	0	1	20	
14448	starting tm alex and i are doing a whole week	neutral	starting tm alex and i are doing a whole week	starting tm alex and i are doing a whole week 	1.000000	21	21	21	
12569	people r weird	negative	r weird	people r weird	0.666667	2	3	3	

5495 rows × 10 columns



bad

In []:

bad = x_val[x_val.jaccard <= 1/3]
bad</pre>

	text	sentiment	pred_text	selected_text	jaccard	Num_words_PT	Num_words_ST	Num_word_text	difference
21278	happy birthday little sister of mine also go	positive	happy birthday	happy birthday little sister of mine	0.333333	2	6	10	
21560	turned my alarm off this morning because i tho	negative	turned my alarm off this morning because i tho	fail	0.043478	26	1	26	
17239	well it almost was a good day guess i just ret	positive	day guess	good day	0.333333	2	2	12	
26658	the ultimate shirt folding tool i saw using	positive	the ultimate shirt folding tool i saw using th	ultimate	0.062500	16	1	16	
2899	oh snap kinda nuts right now ive told at lea	positive	oh snap kinda nuts right now ive told at least	thanks	0.083333	12	1	12	
7554	nice i wish twitter would tile mine	positive	nice i wish twitter	nice	0.250000	4	1	7	
	awww boo i ABUSE		awww boo i						

6135	lo va	sentiment positive	abuse pred text	selected text	o! 26636 8	Num_words_PT	Num_words_St	Num_word_text	difference_
	the girly th		doing all the girly thi						
6816	has a gym day and is hoping to enjoy the last	positive	to enjoy the	enjoy	0.333333	3	1	15	
8829	pappadeux has some yummy strawberry lemonade	positive	##ux has some yu	yummy strawberry lemonade	0.000000	4	3	6	
24019	omg i saw ur update nd it said david archulet	positive		awsome	0.000000	0	1	20	

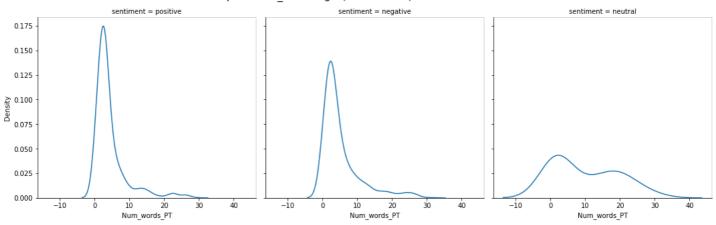
1875 rows × 10 columns

In []:

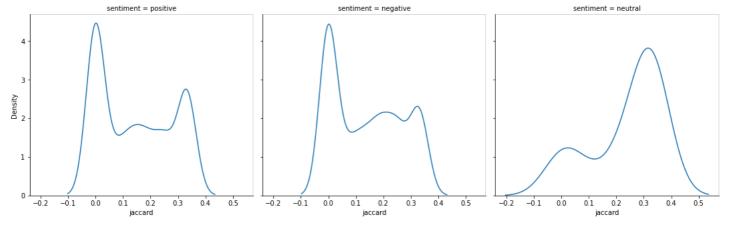
```
g=sns.FacetGrid(data=bad,col='sentiment',height=5);
g.map(sns.kdeplot,'Num_words_PT');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of predicted_text length(no.of words) for various sentiment
labels',fontsize=16);

g=sns.FacetGrid(data=bad,col='sentiment',height=5);
g.map(sns.kdeplot,'jaccard');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of Jaccard score(selected_text & predicted_text) for various
sentiment labels',fontsize=16);
```

Distribution of predicted_text length(no.of words) for various sentiment labels



Distribution of Jaccard score(selected_text & predicted_text) for various sentiment labels



opservations

from the predicted text length curves.we can see that positive and negative texts number of words are more likely between **0 to 20 words** unlike neutral text.

from distributions of jaccard scores of selected_text&predicted_text.here also we can see both positive and negative texts have mostly zero densed values unlike netral text

conclusion

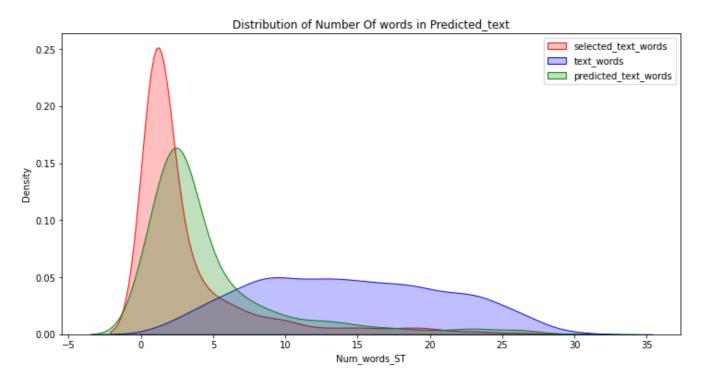
positive and negative texts have zero neared jaccard scores unlike neutral texts

In []:

```
plt.figure(figsize=(12,6))
pl=sns.kdeplot(bad['Num_words_ST'], shade=True, color="r").set_title('Distribution of Num
ber Of words in Predicted_text')
pl=sns.kdeplot(bad['Num_word_text'], shade=True, color="b")
pl=sns.kdeplot(bad['Num_words_PT'], shade=True, color="g")
plt.legend(labels=['selected_text_words','text_words','predicted_text_words'])
```

Out[]:

<matplotlib.legend.Legend at 0x7fb7f06b2510>



observation

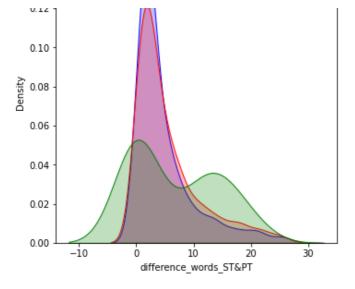
predicted text and selected text are some what similar distributions

In []:

```
plt.figure(figsize=(12,6))
plt.subplot(1,2,1)
pl=sns.kdeplot(bad[bad['sentiment']=='positive']['difference_words_ST&PT'], shade=True,
color="b").set_title('Distribution of Differences btw Selected_text and Predicted_text')
p2=sns.kdeplot(bad[bad['sentiment']=='negative']['difference_words_ST&PT'], shade=True,
color="r")
p2=sns.kdeplot(bad[bad['sentiment']=='neutral']['difference_words_ST&PT'], shade=True, c
olor="g")
plt.legend(labels=['difference in positive_words', 'difference in negative_words', 'difference in neutral_words']);
```

Distribution of Differences btw Selected text and Predicted text





observations

here difference represents predicted_text length > selected_text length

distributions of differences(length) of positive and negative texts are right skewed ranging from 0 to 30 words but for neutral text is different

conclusion

difference words distribution for positive and negative are very overlapping and right skewed

medium

In []:

```
\label{eq:medium} medium = x_val[(x_val.jaccard > 1/3) & (x_val.jaccard <= 2/3)] \\ medium
```

	tevt	sentiment	nred text	selected text	iaccard	Num_words_PT	Num words ST	Num word text	differen
8870	elaines my online mommy too she gives good a	positive	good advice	she gives good advice	0.500000	2	4	9	
6693	there are days of summervac school comes alo	neutral	there are days of summervac school comes along	the annual problem of r generation is finding	0.590909	24	13	24	
20672	time for tv in bedthen spending all day catchi	negative	online summer	i hate online summer classes	0.400000	2	5	17	
2859	soooo happy your back	positive	soooo happy	soooo happy your back	0.500000	2	4	4	
7956	lol its not the army its starfleet rules are	negative	lol its not the army its starfleet rules are m	rules are meant to be broken to get the job done	0.500000	24	11	24	
	extremlv			extremlv			-		

17773	qisesis	neutral sentiment	extremly pred_text	selected_dresit	0.500000 jaccard	Num_words_PT	Num_words_ST	Num_word_text	differen
1892	missing mrs mcfox	negative	missing mrs	missing	0.500000	2	1	3	
25589	my brain is gonna explode a minute from now i	negative	my brain is gonna explode a minute from now it	my brain is gonna explode a minute from now	0.409091	25	9	25	
21533	good one	positive	good	good one	0.500000	1	2	2	
12569	people r weird	negative	r weird	people r weird	0.666667	2	3	3	

917 rows × 10 columns

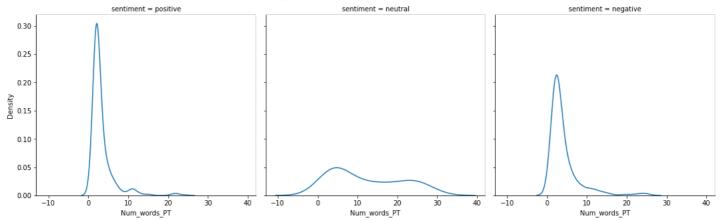
[4]

In []:

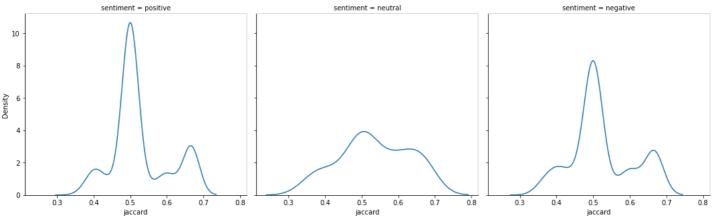
```
g=sns.FacetGrid(data=medium, col='sentiment', height=5);
g.map(sns.kdeplot,'Num_words_PT');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of predicted_text length(no.of words) for various sentiment
labels', fontsize=16);

g=sns.FacetGrid(data=medium, col='sentiment', height=5);
g.map(sns.kdeplot,'jaccard');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of Jaccard score(selected_text & predicted_text) for various
sentiment labels', fontsize=16);
```

Distribution of predicted_text length(no.of words) for various sentiment labels



$Distribution \ of \ Jaccard \ score(selected_text \ \& \ predicted_text) \ for \ various \ sentiment \ labels$



observations

from the predicted text length curves.we can see that positive and negative texts number of words are more likely between 0 to 10 words unlike neutral text.

from distributions of jaccard scores of selected_text&predicted_text.here we can see jaccard scores for each

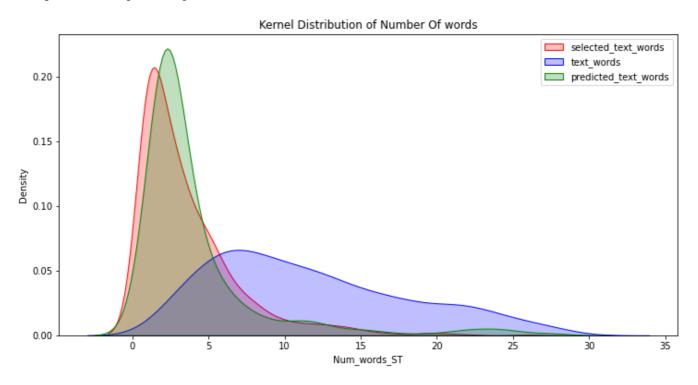
sentiment have mostly 0.5 centered values

In []:

```
plt.figure(figsize=(12,6))
pl=sns.kdeplot(medium['Num_words_ST'], shade=True, color="r").set_title('Kernel Distribu
tion of Number Of words')
pl=sns.kdeplot(medium['Num_word_text'], shade=True, color="b")
pl=sns.kdeplot(medium['Num_words_PT'], shade=True, color="g")
plt.legend(labels=['selected_text_words','text_words','predicted_text_words'])
```

Out[]:

<matplotlib.legend.Legend at 0x7fb6e583e250>

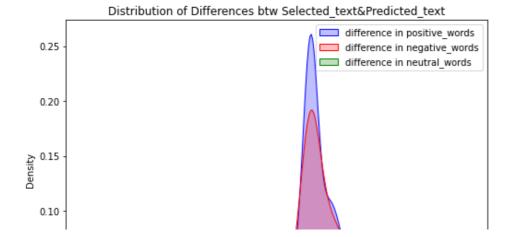


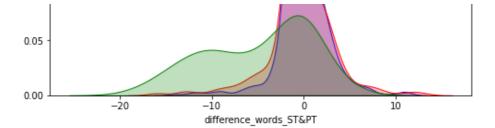
observation

selected text and predicted text have almost similar distributions of lenghts

In []:

```
plt.figure(figsize=(18,6))
plt.subplot(1,2,1)
pl=sns.kdeplot(medium[medium['sentiment']=='positive']['difference_words_ST&PT'], shade=
True, color="b").set_title('Distribution of Differences btw Selected_text&Predicted_text')
p2=sns.kdeplot(medium[medium['sentiment']=='negative']['difference_words_ST&PT'], shade=
True, color="r")
p2=sns.kdeplot(medium[medium['sentiment']=='neutral']['difference_words_ST&PT'], shade=T
rue, color="g")
plt.legend(labels=['difference in positive_words','difference in negative_words','difference in neutral_words']);
```





observations

we see more negative values in neutral text which implies predicted_text of neutral text length > selected_text length

difference words lies between -10 to 10 mostly which is better than bad performed data where it was 0 to 30(wide)

good

In []:

good = x_val[x_val.jaccard > 2/3]
good

shinnity

shinaity

shinnity

	text	sentiment	pred_text	selected_text	jaccard	Num_words_PT	Num_words_ST	Num_word_text	difference
7917	this is my update	neutral	this is my update	this is my update	1.0	4	4	4	
15845	whaaat i still have next week	neutral	whaaat i still have next week	whaaat i still have next week	1.0	6	6	6	
22338	do some research for my article	neutral	do some research for my article	do some research for my article	1.0	6	6	6	
22474	are you okay	neutral	are you okay	are you okay	1.0	3	3	3	
22030	good ABUSE homie hahahahaha thats what im tal	positive	good	good	1.0	1	1	9	
27309	the only thing is im not a chocolate fan but	neutral	the only thing is im not a chocolate fan but t	the only thing is im not a chocolate fan but	1.0	14	14	14	
616	either way you always tend to make my followf	positive	you do rock that much	you do rock that much	1.0	5	5	16	
4505	awww whered you get that hugh is so thin no	neutral	awww whered you get that hugh is so thin no tr	awww whered you get that hugh is so thin no t	1.0	22	22	22	
	gone to the		gone to	gone to the					

15107	shwa b tax it	neutral sentiment	pred <u>s</u> itext	selekted <u>b</u> text	1.0 jaccard	$\begin{array}{c} 7 \\ \text{Num_words_PT} \end{array}$	${\color{red}\text{Num_words_ST}}^{7}$	Num_word_text	difference.
	sunday		back sunday	sunday					
14448	starting tm alex and i are doing a whole week 	neutral	starting tm alex and i are doing a whole week	starting tm alex and i are doing a whole week 	1.0	21	21	21	

2703 rows × 10 columns

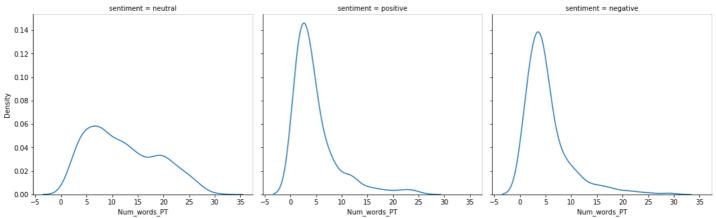
[4]

In []:

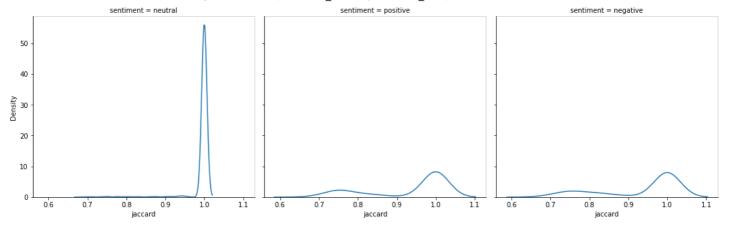
```
g=sns.FacetGrid(data=good, col='sentiment', height=5);
g.map(sns.kdeplot,'Num_words_PT');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of predicted_text length(no.of words) for various sentiment
labels', fontsize=16);

g=sns.FacetGrid(data=good, col='sentiment', height=5);
g.map(sns.kdeplot,'jaccard');
plt.subplots_adjust(top=0.87,)
g.fig.suptitle('Distribution of Jaccard score(selected_text & predicted_text) for various
sentiment labels', fontsize=16);
```

$Distribution\ of\ predicted_text\ length (no. of\ words)\ \ for\ various\ sentiment\ labels$



Distribution of Jaccard score(selected text & predicted text) for various sentiment labels



observations

from the predicted text length curves.we can see that positive and negative texts number of words are more likely between **0 to 10 words** unlike neutral text.

from distributions of jaccard scores of selected_text&predicted_text.here we can see jaccard scores for each sentiment have mostly 1.0 centered values

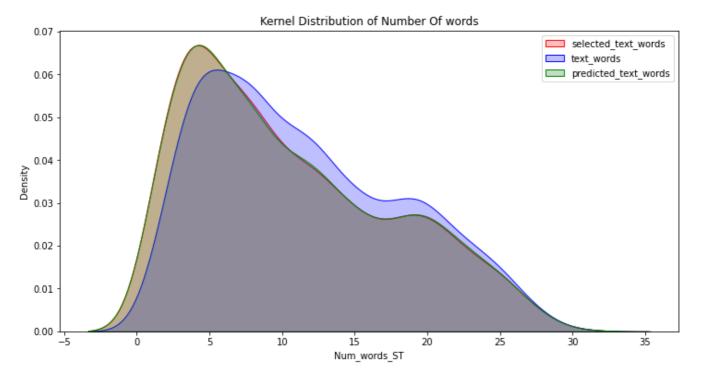
In []:

plt.figure(figsize=(12.6))

```
p1=sns.kdeplot(good['Num_words_ST'], shade=True, color="r").set_title('Distribution of Nu mber Of words')
p1=sns.kdeplot(good['Num_word_text'], shade=True, color="b")
p1=sns.kdeplot(good['Num_words_PT'], shade=True, color="g")
p1=sns.kdeplot(good['Num_words_PT'], shade=True, color="g")
p1=sns.kdeplot(good['Num_words_PT'], shade=True, color="g")
```

Out[]:

<matplotlib.legend.Legend at 0x7fb6e57b9d50>

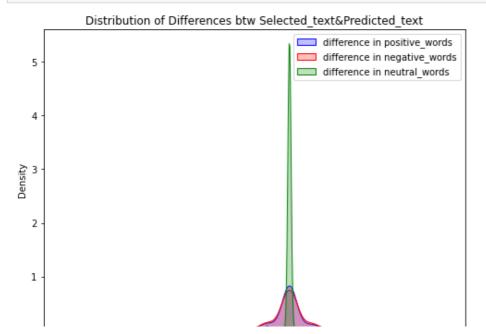


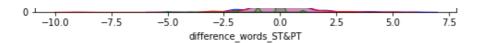
observation

selected text and predicted text have exact same distribution of lenghts

In []:

```
plt.figure(figsize=(18,6))
plt.subplot(1,2,1)
pl=sns.kdeplot(good[good['sentiment']=='positive']['difference_words_ST&PT'], shade=True
, color="b").set_title('Distribution of Differences btw Selected_text&Predicted_text')
p2=sns.kdeplot(good[good['sentiment']=='negative']['difference_words_ST&PT'], shade=True
, color="r")
p2=sns.kdeplot(good[good['sentiment']=='neutral']['difference_words_ST&PT'], shade=True,
color="g")
plt.legend(labels=['difference in positive_words','difference in negative_words','difference in neutral_words']);
```





observations

difference in words is zero centered for all sentiments

positive and negative sentiments have same kind of distribution and lies between -2 to 2 words

by densities of each sentiment. we can say that neutral texts are more

conclusion

difference words are few so jaccard performance is good

this data is good because of presence of neutral text

final conclusion

we can not say length of text decides jaccard performance because same selected_text length of 0-20words we are getting 0 to 1 jaccard scores for bad,medium,good data respectively.