In this notebook, You will do amazon review classification with BERT. It contains 5 parts as below. Detailed instrctions are given in the each cell. please read every comment we have written.

- 1. Preprocessing
- 2. Creating a BERT model from the Tensorflow HUB.
- 3. Tokenization
- 4. getting the pretrained embedding Vector for a given review from the BERT.
 - 5. Using the embedding data apply NN and classify the reviews.

instructions:

- 1. Don't change any Grader Functions. Don't manipulate any Grader functions. If you manipulate any, it will be considered as plagiarised.
- 2. Please read the instructions on the code cells and markdown cell s. We will explain what to write.
- 3. please return outputs in the same format what we asked. Eg. Don't return List of we are asking for a numpy array.
- 4. Please read the external links that we are given so that you will learn the concept behind the code that you are writing.
- 5. We are giving instructions at each section if necessary, please f ollow them.

Every Grader function has to return True.

Grader function 1

1 #Read the dataset - Amazon fine food reviews

In []:

```
$
```

```
3 #check the info of the dataset
          4 reviews.info()
In [ ]:
          1 #get only 2 columns - Text, Score
          2 #drop the NAN values
In [ ]:
          1 #if score> 3, set score = 1
          2 #if score<=2, set score = 0
          3 #if score == 3, remove the rows.
        Grader function 2
In [ ]:
            def grader reviews():
                 temp shape = (reviews.shape == (525814, 2)) and (reviews.Score.value cou
          2
          3
                 assert(temp shape == True)
          4
                 return True
          5 grader reviews()
In [ ]:
         1 def get_wordlen(x):
          2
                 return len(x.split())
          3 reviews['len'] = reviews.Text.apply(get wordlen)
          4 reviews = reviews[reviews.len<50]</pre>
          5 reviews = reviews.sample(n=100000, random_state=30)
In [ ]:
          1 #remove HTML from the Text column and save in the Text column only
In [ ]:
          1 #print head 5
In [ ]:
          1 | #split the data into train and valudation data(20%) with Stratify sampling,
In [ ]:
          1 #plot bar graphs of y_train and y_test
In [ ]:
          1 #saving to disk. if we need, we can load preprocessed data directly.
          2 reviews.to_csv('preprocessed.csv', index=False)
```

2 reviews = pd.read csv(r"D:\ML\Internal DL\NLP\amazon-fine-food-reviews\Revie

Part-2: Creating BERT Model

If you want to know more about BERT, You can watch live sessions on Tran sformers and BERt. we will strongly recommend you to read Transformers
https://jalammar.github.io/illustrated-transformer/), BERT Paper (https://jalammar.github.i

For this assignment, we are using <u>BERT uncased Base model (https://tfhub.dev/tensorflow/bert_en_uncased_L-12_H-768_A-12/1)</u>. It uses L=12 hidden layers (i.e., Transformer blocks), a hidden size of H=768, and A=12 attention heads.

```
In [ ]:
            ## Loading the Pretrained Model from tensorflow HUB
          2 tf.keras.backend.clear session()
          3
          4 # maximum length of a seg in the data we have, for now i am making it as 300
          5 max seq length = 55
          6
          7
            #BERT takes 3 inputs
          8
          9
           #this is input words. Sequence of words represented as integers
            input word ids = tf.keras.layers.Input(shape=(max seq length,), dtype=tf.int
         10
         11
         12 #mask vector if you are padding anything
         input_mask = tf.keras.layers.Input(shape=(max_seq_length,), dtype=tf.int32,
         14
         15 #segment vectors. If you are giving only one sentence for the classification
         16 #If you are giving two sentenced with [sep] token separated, first seg segme
         17 #second seg segment vector are 1's
         18 segment ids = tf.keras.layers.Input(shape=(max seq length,), dtype=tf.int32,
         19
         20 #bert Laver
         21 bert layer = hub.KerasLayer("https://tfhub.dev/tensorflow/bert en uncased L-
            pooled output, sequence output = bert layer([input word ids, input mask, seg
         22
         23
         24 #Bert model
         25 #We are using only pooled output not sequence out.
         26 #If you want to know about those, please read https://www.kagqle.com/questio
         27
            bert model = Model(inputs=[input word ids, input mask, segment ids], outputs
         28
In [ ]:
          1 bert model.summary()
In [ ]:
            bert model.output
```

Part-3: Tokenization

```
In [ ]: | 1 | #import tokenization - We have given tokenization.py file
```

Grader function 3

#it has to give no error

In []:

```
def grader tokenize(tokenizer):
          2
          3
                 out = False
                 try:
          4
          5
                     out=('[CLS]' in tokenizer.vocab) and ('[SEP]' in tokenizer.vocab)
          6
                 except:
          7
                     out = False
          8
                 assert(out==True)
          9
                 return out
         10 grader_tokenize(tokenizer)
In [ ]:
            # Create train and test tokens (X_train_tokens, X_test_tokens) from (X_train_
          1
          2
          3 # add '[CLS]' at start of the Tokens and '[SEP]' at the end of the tokens.
            # maximum number of tokens is 55(We already given this to BERT layer above)
          5
          6
            # if it is less than 55, add '[PAD]' token else truncate the tokens length.(
          7
          8
          9
            # Based on padding, create the mask for Train and Test ( 1 for real token, 0
         10 # it will also same shape as input tokens (None, 55) save those in X train m
         11
         12
            # Create a segment input for train and test. We are using only one sentence
         13
         14 # type of all the above arrays should be numpy arrays
         15
         16 # after execution of this cell, you have to get
         17 | # X train tokens, X train mask, X train segment
         18 # X_test_tokens, X_test_mask, X_test_segment
```

Example

```
3 print('='*50)
             4 tokens = tokenizer.tokenize(X_train.values[0])
             5 # we need to do this "tokens = tokens[0:(max_seq_length-2)]" only when our len(tokens) is more than "max_seq_length - 2"
             6 # we will consider only the tokens from 0 to max_seq_length-2
             7 # if our len(tokens) are < max_seq_length-2, we don't need to do this
             8 tokens = tokens[0:(max_seq_length-2)]
             9 # we are doing that so that we can include the tokens [CLS] and [SEP] and make the whole sequence length == max_seq_length
             10 tokens = ['[CLS]',*tokens,'[SEP]']
             11 print("tokens are: \n", np.array(tokens))
             12 print('='*50)
            13 print("number of tokens :",len(tokens))
            14 print("tokens replaced with the positional encoding :\n",np.array(tokenizer.convert_tokens_to_ids(tokens)))
            15 print('='*50)
            16 print("the mask array is : ", np.array([1]*len(tokens)+[0]*(max_seq_length-len(tokens))))
            18 print("the segment array is :",np.array([0]*max_seq_length))
            19 print('='*50)
            original sentance :
             ['I' 'had' 'never' 'tried' 'this' 'brand' 'before,' 'so' 'I' 'was' 'worried' 'about' 'the' 'quality.' 'It' 'tasted' 'great.' 'A' 'very' 'nice' 'smooth' 'rich' 'full' 'flavor.' 'Its' 'my' 'new' 'favoret.']
                                                         'new' 'favoret.']
            number of words: 28
             ['[CLS]' 'i' 'had' 'never' 'tried' 'this' 'brand' 'before' ',' 'so' 'i'
'was' 'worried' 'about' 'the' 'quality' '.' 'it' 'tasted' 'great' '.' 'a'
'very' 'nice' 'smooth' 'rich' 'full' 'flavor' '.' 'its' 'my' 'new'
'favor' '##et' '.' '[SEP]']
             _____
            number of tokens : 36
            tokens replaced with the positional encoding :
             [ 101 1045 2018 2196 2699 2023 4435 2077 1010 2061 1045 2001
              5191 2055 1996 3737 1012 2009 12595 2307 1012 1037 2200 3835
              5744 4138 2440 14894 1012 2049 2026 2047 5684 3388 1012 102]
             _____
            In [ ]:
              1 | import pickle
In [ ]:
              1 | ##save all your results to disk so that, no need to run all again.
              2 | pickle.dump((X_train, X_train_tokens, X_train_mask, X_train_segment, y_train
              3 | pickle.dump((X_test, X_test_tokens, X_test_mask, X_test_segment, y_test),ope
In [ ]:
             1 #you can load from disk
              2 | #X train, X train tokens, X train mask, X train segment, y train = pickle.lo
              3 |#X_test, X_test_tokens, X_test_mask, X_test_segment, y_test = pickle.load(op)
```

1 print("original sentance : \n", np.array(X_train.values[0].split()))
2 print("number of words: ", len(X_train.values[0].split()))

Grader function 4

```
In [ ]:
             def grader_alltokens_train():
          1
          2
                 out = False
          3
          4
                 if type(X train tokens) == np.ndarray:
          5
          6
                     temp_shapes = (X_train_tokens.shape[1]==max_seq_length) and (X_train_
          7
                     (X_train_segment.shape[1]==max_seq_length)
          8
          9
                     segment_temp = not np.any(X_train_segment)
         10
         11
                     mask_temp = np.sum(X_train_mask==0) == np.sum(X_train_tokens==0)
         12
                     no_cls = np.sum(X_train_tokens==tokenizer.vocab['[CLS]'])==X_train_t
         13
         14
                     no_sep = np.sum(X_train_tokens==tokenizer.vocab['[SEP]'])==X_train_t
         15
         16
         17
                     out = temp_shapes and segment_temp and mask_temp and no_cls and no_s
         18
         19
                 else:
         20
                     print('Type of all above token arrays should be list not numpy array
         21
                     out = False
         22
                 assert(out==True)
         23
                 return out
         24
         25
             grader_alltokens_train()
```

Grader function 5

```
In [ ]:
             def grader_alltokens_test():
          1
          2
                 out = False
          3
                 if type(X_test_tokens) == np.ndarray:
          4
          5
                     temp_shapes = (X_test_tokens.shape[1]==max_seq_length) and (X_test_m
                     (X_test_segment.shape[1]==max_seq_length)
          6
          7
          8
                     segment_temp = not np.any(X_test_segment)
          9
         10
                     mask temp = np.sum(X test mask==0) == np.sum(X test tokens==0)
         11
         12
                     no_cls = np.sum(X_test_tokens==tokenizer.vocab['[CLS]'])==X_test_tok
         13
         14
                     no_sep = np.sum(X_test_tokens==tokenizer.vocab['[SEP]'])==X_test_tok
         15
         16
                     out = temp shapes and segment temp and mask temp and no cls and no s
         17
         18
                 else:
         19
                     print('Type of all above token arrays should be list not numpy array
         20
                     out = False
         21
                 assert(out==True)
         22
                 return out
         23
            grader_alltokens_test()
```

Part-4: Getting Embeddings from BERT M

odel

We already created the BERT model in the part-2 and input data in the part-3. We will utlize those two and will get the embeddings for each sent ence in the Train and Validation data.

```
In []: 1 bert_model.input

In []: 1 bert_model.output

In []: 1 # get the train output, BERT model will give one output so save in 2 # X_train_pooled_output

In []: 1 # get the test output, BERT model will give one output so save in 2 # X_test_pooled_output

In []: 1 ##save all your results to disk so that, no need to run all again. 2 pickle.dump((X_train_pooled_output, X_test_pooled_output),open('final_output)

In []: 1 #X_train_pooled_output, X_test_pooled_output= pickle.load(open('final_output))
```

Grader function 6

```
In [ ]:
            #now we have X train pooled output, y train
          2 #X_test_pooled_ouput, y_test
          3
            #please use this grader to evaluate
          4
          5
            def greader_output():
                 assert(X_train_pooled_output.shape[1]==768)
          6
          7
                 assert(len(y train)==len(X train pooled output))
          8
                 assert(X test pooled output.shape[1]==768)
          9
                 assert(len(y_test)==len(X_test_pooled_output))
         10
                 assert(len(y train.shape)==1)
         11
                 assert(len(X_train_pooled_output.shape)==2)
         12
                 assert(len(y_test.shape)==1)
         13
                 assert(len(X test pooled output.shape)==2)
         14
                 return True
         15 greader_output()
```

Part-5: Training a NN with 786 feature s

Create a NN and train the NN.

- 1. You have to use AUC as metric.
- 2. You can use any architecture you want.
- 3. You have to use tensorboard to log all your metrics and Losses. You have to send those logs.

- 4. Print the loss and metric at every epoch.
- 5. You have to submit without overfitting and underfitting.