BERT

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From: SVNIT, Gujarat

Mtech Data Science - p23ds004 (2023-25)

Subject: NLP Project

Last Updated:

In [1]: import pandas as pd
import numpy as np

In [2]: data = pd.read_csv("dataset.csv")

In [3]: data

Out[3]:

	tweets	class	target	bert_vectors
0	Be aware dirty step to get money #staylight	figurative	0.0	[0.5142972 0.09474187 0.41039863 -0.086846
1	#sarcasm for #people who don't understand #diy	figurative	0.0	[0.60709447 0.24507785 0.30069906 -0.095903
2	@IminworkJeremy @medsingle #DailyMail readers	figurative	0.0	[0.12725917 0.07618354
3	@wilw Why do I get the feeling you like games?	figurative	0.0	[4.73182350e-01 5.66805065e-01 3.70653301e
4	-@TeacherArthurG @rweingarten You probably jus	figurative	0.0	[0.50476366 0.16474849 0.34438178 0.067346
81403	Photo: Image via We Heart It http://t.co/ky8Nf	sarcasm	1.0	[2.27706909e-01 -4.00034375e-02 5.59347928e
81404	I never knewI better put this out to the Uni	sarcasm	1.0	[0.21134636 0.16012657 0.7607299 -0.195773
81405	hey just wanted to say thanks @ puberty for le	sarcasm	1.0	[0.39612132 0.26797482
81406	I'm sure coverage like the Fox News Special "T	sarcasm	1.0	[0.35054675 -0.17652354
81407	@skeyno16 at u13?! I won't believe it until I	sarcasm	1.0	[0.34130692 0.04276937 0.69224215 -0.030625

81408 rows × 4 columns

```
data['bert_vectors']
In [4]:
Out[4]: 0
                0.5142972
                              0.09474187
                                         0.41039863 -0.086846...
        1
                [ 0.60709447
                              0.24507785
                                         0.30069906 -0.095903...
        2
                  0.12725917
                             0.07618354
                                         0.5803481
                                                     0.085881...
        3
                [ 4.73182350e-01 5.66805065e-01 3.70653301e-...
        4
                [ 0.50476366  0.16474849  0.34438178  0.067346...
        81403
                [ 2.27706909e-01 -4.00034375e-02 5.59347928e-...
        81404
                [ 0.21134636  0.16012657
                                         0.7607299 -0.195773...
        81405
                [ 0.39612132  0.26797482
                                         0.8456364 -0.222941...
        81406
                [ 0.35054675 -0.17652354
                                         0.36843812 -0.002357...
        81407
                [ 0.34130692  0.04276937
                                         0.69224215 -0.030625...
        Name: bert_vectors, Length: 81408, dtype: object
In [5]: |data['bert_vectors'][0]
Out[5]: '[ 0.5142972
                      0.09474187   0.41039863   -0.08684609   0.2877095
        \n 0.40892476 0.44171068 -0.01723384 -0.26785442 -0.16827619 -0.22784133
        \n -0.31953806
                       0.275655
                                  -0.10393742 0.49432114 -0.02620821
                                                                    0.2063558
        \n -0.23674724  0.25816718  0.34353077
                                              0.0214939 -0.13152978 0.11771113
        \n 0.49273565 -0.15442434 0.10735898 -0.15738115 -0.3916646 -0.16478181
                                             \n 0.49133274 -0.04270812
                                  0.237843
        \n -0.31194413 -0.0389487
                                  0.29511118
                                              0.44777897 -0.36570454 -0.5686758
        \n 0.19408512 0.02260724 -0.31137243 -0.28884372 0.2681975
                                                                     0.26582816
        \n 0.33216658 -0.15465686 -0.2563196
                                              0.3087412
                                                        -0.09471447 -0.08942274
        \n 0.0777904
                       0.74242663
                                  0.3745379
                                             -0.45100695 -0.43329164 -0.02466331
        \n -0.16693549  0.08048202  0.26112032  -0.30901015  0.12328291  0.21376407
                       0.6823772 -1.0520166
                                              0.28137955 -0.01561841 -0.03059365
        \n -0.0388822
        \n -0.06478383
                       0.13605814 0.01955121
                                              0.20722364 -0.10848905
                                                                     0.6780405
        \n 0.31807926 -0.0370811
                                  0.05020353
                                              0.16420896 0.10430315
                                                                     0.38346\n
        0.05057348 0.08532551 0.19232303 0.20832938 0.40885463 -0.42770848\n
                              -0.24444808 0.3981791
        0.01001668 -0.3563449
                                                      0.30680826 -0.26975936\n -
        0.02268203 0.10180932 -0.46551192
                                         0.16729726 -0.01927103 -0.12172183\n
        0.12085328 0.32595316
                              0.14308026
                                         0.17479685 -0.05908208 -0.4421207\n
        0.24812964 0.16455555
                              0.15420035
                                          0.206105
                                                      0.23889849
                                                                 0.03915282\n -
        0.02392515
                   0.22758889
                               0.21866032
                                          0.78823656
                                                      0.20028736
                                                                 0.3199013\n -
                                          0.12637198 0.41490802
        0.40904292 0.2744289
                               0.16341347
                                                                 0.29418418\n -
        0.08437694 - 0.54106766 - 0.11171681 - 0.07831711 - 0.02037072 - 0.10411447 \ 
        0.11889234 -0.14834784 -0.08240354 -0.39731655
                                                      0.12649344
                                                                 0.1883826\n
        0.09964897 -0.01768736 0.12272178 0.03364825 -0.17410624
                                                                 0.5015001 ]'
In [6]: data['class'].value_counts()
Out[6]: class
        figurative
                     21238
        irony
                     20894
        sarcasm
                     20681
        regular
                     18595
        Name: count, dtype: int64
In [7]:
        print(type(data['bert_vectors'][0]))
        <class 'str'>
```

```
In [8]: def str_to_vec(input_string):
               # Your input string
               # input_string = "[ 0.47350284 0.17998783 0.52607228 -0.21903914 0.5047
               # Remove brackets and newline characters
               cleaned_string = input_string.replace('[', '').replace(']', '').replace
               # Split the string into a list of strings
               string_values = cleaned_string.split()
               # Convert each string value to a float
               float_values = [float(value) for value in string_values]
               # Convert the list of floats to a NumPy array
               vector = np.array(float_values)
               # Now, 'vector' is a NumPy array representing your vector
               return vector
         dataset = pd.DataFrame()
 In [9]:
          dataset['bert_vectors'] = data['bert_vectors'].apply(str_to_vec)
In [11]:
          dataset
Out[11]:
                                                bert_vectors
               0 [0.5142972, 0.09474187, 0.41039863, -0.0868460...
               1 [0.60709447, 0.24507785, 0.30069906, -0.095903...
               2 [0.12725917, 0.07618354, 0.5803481, 0.0858818,...
               3 [0.47318235, 0.566805065, 0.370653301, -0.0819...
               4 [0.50476366, 0.16474849, 0.34438178, 0.0673463...
           81403
                  [0.227706909, -0.0400034375, 0.559347928, -0.1...
           81404
                  [0.21134636, 0.16012657, 0.7607299, -0.1957732...
           81405
                  [0.39612132, 0.26797482, 0.8456364, -0.2229413...
           81406
                  [0.35054675, -0.17652354, 0.36843812, -0.00235...
                  [0.34130692, 0.04276937, 0.69224215, -0.030625...
           81407
          81408 rows × 1 columns
In [12]:
          print(len(data['bert_vectors']))
          81408
```

```
dataset['bert_vectors'][0]
In [13]:
Out[13]: array([ 0.5142972 ,
                             0.09474187, 0.41039863, -0.08684609,
                                                                   0.2877095 ,
                -0.16383709, 0.40892476, 0.44171068, -0.01723384, -0.26785442,
                -0.16827619, -0.22784133, -0.31953806, 0.275655 , -0.10393742,
                 0.49432114, -0.02620821, 0.2063558, -0.23674724,
                                                                  0.25816718,
                 0.34353077, 0.0214939, -0.13152978, 0.11771113, 0.49273565,
                -0.15442434, 0.10735898, -0.15738115, -0.3916646 , -0.16478181,
                 0.49133274, -0.04270812, 0.237843 , -0.20905466, 0.06037405,
                -0.17967163, -0.31194413, -0.0389487, 0.29511118, 0.44777897,
                -0.36570454, -0.5686758, 0.19408512, 0.02260724, -0.31137243,
                             0.2681975, 0.26582816, 0.33216658, -0.15465686,
                -0.28884372,
                -0.2563196 , 0.3087412 , -0.09471447 , -0.08942274 , 0.0777904 ,
                             0.3745379 , -0.45100695, -0.43329164, -0.02466331,
                -0.16693549, 0.08048202, 0.26112032, -0.30901015,
                                                                   0.12328291,
                 0.21376407, -0.0388822 , 0.6823772 , -1.0520166 ,
                                                                   0.28137955,
                -0.01561841, -0.03059365, -0.06478383, 0.13605814, 0.01955121,
                 0.20722364, -0.10848905, 0.6780405, 0.31807926, -0.0370811,
                 0.05020353, 0.16420896, 0.10430315, 0.38346
                                                                   0.19076318,
                             0.11541953, -0.00839414, -0.60080236,
                 0.27409777,
                                                                   0.47035944,
                -0.05057348, 0.08532551, 0.19232303, 0.20832938, 0.40885463,
                -0.42770848, 0.01001668, -0.3563449, -0.24444808,
                                                                   0.3981791 ,
                 0.30680826, -0.26975936, -0.02268203, 0.10180932, -0.46551192,
                 0.16729726, -0.01927103, -0.12172183, 0.12085328, 0.32595316,
                 0.14308026, 0.17479685, -0.05908208, -0.4421207,
                                                                   0.24812964,
                 0.16455555, 0.15420035, 0.206105 , 0.23889849,
                                                                   0.03915282,
                             0.22758889, 0.21866032,
                -0.02392515,
                                                      0.78823656,
                                                                   0.20028736,
                 0.3199013 , -0.40904292 , 0.2744289 , 0.16341347 , 0.12637198 ,
                 0.41490802, 0.29418418, -0.08437694, -0.54106766, -0.11171681,
                -0.07831711, -0.02037072, -0.10411447, -0.11889234, -0.14834784,
                -0.08240354, -0.39731655, 0.12649344, 0.1883826, 0.09964897,
                -0.01768736, 0.12272178, 0.03364825, -0.17410624, 0.5015001 ])
```

Create Input and Output

```
In [14]: def create_input_vectors(vector):
    n = len(vector)
    array_2d = np.empty((n, len(vector[0])), dtype=object)

# Create and insert 10 random 5D arrays into the 2D array
for i in range(n):
    array_150d = vector[i]
    array_2d[i] = array_150d

    return array_2d

In [15]: input_vec = create_input_vectors(dataset['bert_vectors'])

In [16]: print(type(input_vec))
    <class 'numpy.ndarray'>
```

```
In [17]: |print(input_vec.ndim)
         2
In [18]: |mapping_dict = {'figurative': 0, 'irony': 1, 'sarcasm': 1, 'regular': 0}
         # Use map to replace string values with numerical values
         output = data['class'].map(mapping_dict).to_numpy()
In [19]: | print(output.ndim)
         1
In [20]: print(type(output))
         <class 'numpy.ndarray'>
In [21]: output
Out[21]: array([0, 0, 0, ..., 1, 1, 1], dtype=int64)
         Data Splitting
In [22]: def split_data(array_2d, ranges_to_copy):
             copied_ranges = []
             # Loop through each range and copy the corresponding elements
             for start, end in ranges_to_copy:
                 copied_range = array_2d[start:end+1] # Adjust end index to include
                 copied_ranges.append(copied_range)
             # Concatenate the copied ranges along the first axis to create the fina
             copied_array = np.concatenate(copied_ranges, axis=0)
             return copied_array
In [23]: x train = split data(input vec, [(0, 16989), (21238, 37952), (42132, 57007)]
         x_test = split_data(input_vec, [(16990, 21237), (37953, 42131), (57008, 607
In [24]: print("x train:", len(x_train))
         print("x test:", len(x_test))
         print("Total:", len(x_train) + len(x_test))
         x train: 65125
         x test: 16283
         Total: 81408
In [25]: y_train = np.concatenate((np.zeros(16990), np.ones(31591), np.zeros(16544))
         y_test = np.concatenate((np.zeros(4248), np.ones(7898), np.zeros(4137)))
```

```
In [26]: print("train:", len(y_train))
print("test:", len(y_test))
print("total:", len(y_train) + len(y_test))
```

train: 65125
test: 16283
total: 81408

Training With Neural Network

```
In [27]: import tensorflow as tf
from tensorflow import keras
```

WARNING:tensorflow:From C:\Users\Harsh Bari\AppData\Local\Programs\Python \Python310\lib\site-packages\keras\src\losses.py:2976: The name tf.losses. sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losse s.sparse_softmax_cross_entropy instead.

Neural Network for Average Word Embedding

WARNING:tensorflow:From C:\Users\Harsh Bari\AppData\Local\Programs\Python \Python310\lib\site-packages\keras\src\backend.py:873: The name tf.get_def ault_graph is deprecated. Please use tf.compat.v1.get_default_graph instea d.

WARNING:tensorflow:From C:\Users\Harsh Bari\AppData\Local\Programs\Python \Python310\lib\site-packages\keras\src\optimizers__init__.py:309: The nam e tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimize r instead.

Check Model Summary

In [29]: awe.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 256)	38656
dense_1 (Dense)	(None, 128)	32896
dense_2 (Dense)	(None, 64)	8256
dense_3 (Dense)	(None, 32)	2080
dense_4 (Dense)	(None, 16)	528
dense_5 (Dense)	(None, 8)	136
dense_6 (Dense)	(None, 2)	18

Total params: 82570 (322.54 KB) Trainable params: 82570 (322.54 KB) Non-trainable params: 0 (0.00 Byte)

Train Model

In [30]: awe.fit(x_train.astype(np.float32), y_train.astype(np.float32), epochs=22)

Epoch 1/22

WARNING:tensorflow:From C:\Users\Harsh Bari\AppData\Local\Programs\Python \Python310\lib\site-packages\keras\src\utils\tf_utils.py:492: The name tf. ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.Rag gedTensorValue instead.

WARNING:tensorflow:From C:\Users\Harsh Bari\AppData\Local\Programs\Python \Python310\lib\site-packages\keras\src\engine\base_layer_utils.py:384: The name tf.executing_eagerly_outside_functions is deprecated. Please use tf.c ompat.v1.executing_eagerly_outside_functions instead.

```
2036/2036 [============= ] - 9s 3ms/step - loss: 0.4457 -
accuracy: 0.7695
Epoch 2/22
2036/2036 [============= ] - 7s 4ms/step - loss: 0.4040 -
accuracy: 0.7932
Epoch 3/22
accuracy: 0.7989
Epoch 4/22
2036/2036 [=============== ] - 7s 4ms/step - loss: 0.3812 -
accuracy: 0.8031
Epoch 5/22
accuracy: 0.8069
Epoch 6/22
2036/2036 [============= ] - 7s 3ms/step - loss: 0.3664 -
accuracy: 0.8109
Epoch 7/22
2036/2036 [============= ] - 7s 3ms/step - loss: 0.3597 -
accuracy: 0.8131
Epoch 8/22
2036/2036 [============= ] - 7s 4ms/step - loss: 0.3552 -
accuracy: 0.8154
Epoch 9/22
2036/2036 [============= ] - 7s 3ms/step - loss: 0.3509 -
accuracy: 0.8165
Epoch 10/22
accuracy: 0.8195
Epoch 11/22
2036/2036 [============= ] - 7s 3ms/step - loss: 0.3401 -
accuracy: 0.8212
Epoch 12/22
accuracy: 0.8237
Epoch 13/22
accuracy: 0.8258
Epoch 14/22
accuracy: 0.8286
Epoch 15/22
accuracy: 0.8302
Epoch 16/22
2036/2036 [============= ] - 7s 3ms/step - loss: 0.3197 -
accuracy: 0.8313
Epoch 17/22
2036/2036 [============== ] - 7s 3ms/step - loss: 0.3164 -
accuracy: 0.8338
```

Out[30]: <keras.src.callbacks.History at 0x152a6742e90>

Training Accuracy

Testing Accuracy

In [35]: print(classification_report(y_test.astype(np.float32), prediction))
 print()
 print("Confusion Matrix: \n", confusion_matrix(y_test.astype(np.float32), p
 print("\nAccuracy: \n", accuracy_score(y_test.astype(np.float32), prediction)

	precision	recall	f1-score	support
0.0	0.88	0.76	0.81	8385
1.0	0.77	0.90	0.83	7898
accuracy			0.82	16283
macro avg	0.83	0.83	0.82	16283
weighted avg	0.83	0.82	0.82	16283

Confusion Matrix: [[6331 2054] [827 7071]]

Accuracy:

0.8230670023951361