▼ Identify the Sentiments

Sentiment analysis is contextual mining of text which identifies and extracts subjective information in source material, and helping a business to understand the social sentiment of their brand, product or service while monitoring online conversations. Brands can use this data to measure the success of their products in an objective manner. In this challenge, you are provided with tweet data to predict sentiment on electronic products of netizens.

```
#Import Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import re
import pickle
%matplotlib inline
import subprocess
!pip install bert-serving-client
!pip install -U bert-serving-server[http]
      Collecting bert-serving-client
        Downloading <a href="https://files.pythonhosted.org/packages/1f/09/aae1405378a848b2e87769ad">https://files.pythonhosted.org/packages/1f/09/aae1405378a848b2e87769ad</a>
      Requirement already satisfied: pyzmq>=17.1.0 in /usr/local/lib/python3.6/dist-package
      Requirement already satisfied: numpy in /usr/local/lib/python3.6/dist-packages (from
      Installing collected packages: bert-serving-client
      Successfully installed bert-serving-client-1.10.0
      Collecting bert-serving-server[http]
        Downloading <a href="https://files.pythonhosted.org/packages/b0/bd/cab677bbd0c5fb08b72e46837">https://files.pythonhosted.org/packages/b0/bd/cab677bbd0c5fb08b72e46837</a>
                                                      71kB 3.9MB/s
      Requirement already satisfied, skipping upgrade: numpy in /usr/local/lib/python3.6/di
      Collecting GPUtil>=1.3.0
        Downloading <a href="https://files.pythonhosted.org/packages/ed/0e/5c61eedde9f6c87713e89d79/">https://files.pythonhosted.org/packages/ed/0e/5c61eedde9f6c87713e89d79/</a>
      Requirement already satisfied, skipping upgrade: six in /usr/local/lib/python3.6/dist
      Requirement already satisfied, skipping upgrade: pyzmq>=17.1.0 in /usr/local/lib/pyth
      Requirement already satisfied, skipping upgrade: termcolor>=1.1 in /usr/local/lib/pyt
      Collecting flask-json; extra == "http"
        Downloading <a href="https://files.pythonhosted.org/packages/6f/2d/4c21d98b11f3a206fabbdd965">https://files.pythonhosted.org/packages/6f/2d/4c21d98b11f3a206fabbdd965</a>
      Collecting flask-cors; extra == "http"
        Downloading <a href="https://files.pythonhosted.org/packages/69/7f/d0aeaaaafb5c3c76c8d2141dbe">https://files.pythonhosted.org/packages/69/7f/d0aeaaaafb5c3c76c8d2141dbe</a>
      Requirement already satisfied, skipping upgrade: flask; extra == "http" in /usr/local
      Requirement already satisfied, skipping upgrade: bert-serving-client; extra == "http'
      Collecting flask-compress; extra == "http"
        Downloading <a href="https://files.pythonhosted.org/packages/de/eb/6bb0f8cb872167752eab8b06">https://files.pythonhosted.org/packages/de/eb/6bb0f8cb872167752eab8b06</a>
      Requirement already satisfied, skipping upgrade: Jinja2>=2.10.1 in /usr/local/lib/pyt
      Requirement already satisfied, skipping upgrade: click>=5.1 in /usr/local/lib/python?
      Requirement already satisfied, skipping upgrade: itsdangerous>=0.24 in /usr/local/lik
      Requirement already satisfied, skipping upgrade: Werkzeug>=0.15 in /usr/local/lib/pyt
      Collecting brotli
        Downloading <a href="https://files.pythonhosted.org/packages/b4/d3/7c98f05b7b9103e2f3a112ba/">https://files.pythonhosted.org/packages/b4/d3/7c98f05b7b9103e2f3a112ba/</a>
                                                   358kB 16.6MB/s
      Requirement already satisfied, skipping upgrade: MarkupSafe>=0.23 in /usr/local/lib/r
      Building wheels for collected packages: GPUtil, flask-compress
```

Building wheel for GPUtil (setup.py) ... done

```
Stored in directory: /root/.cache/pip/wheels/3d/77/07/80562de4bb0786e5ea186911a2c8
       Building wheel for flask-compress (setup.py) ... done
       Created wheel for flask-compress: filename=Flask Compress-1.7.0-cp36-none-any.whl s
       Stored in directory: /root/.cache/pip/wheels/1d/b7/18/2b88ed33c5ef53868d1bfb0d3f2f3
     Successfully built GPUtil flask-compress
     Installing collected packages: GPUtil, flask-json, flask-cors, brotli, flask-compress
     Successfully installed GPUtil-1.4.0 bert-serving-server-1.10.0 brotli-1.0.9 flask-com
    4
# Download and unzip the pre-trained model
!wget http://storage.googleapis.com/bert_models/2018_10_18/uncased_L-12_H-768_A-12.zip
!unzip uncased_L-12_H-768_A-12.zip
     --2020-10-26 20:40:02-- <a href="http://storage.googleapis.com/bert">http://storage.googleapis.com/bert</a> models/2018 10 18/uncase(
     Resolving storage.googleapis.com (storage.googleapis.com)... 74.125.195.128, 74.125.1
     Connecting to storage.googleapis.com (storage.googleapis.com) 74.125.195.128 : 80... (
     HTTP request sent, awaiting response... 200 OK
     Length: 407727028 (389M) [application/zip]
     Saving to: 'uncased_L-12_H-768_A-12.zip'
     uncased_L-12_H-768_ 100%[=========>] 388.84M 120MB/s
                                                                           in 3.2s
     2020-10-26 20:40:06 (120 MB/s) - 'uncased_L-12_H-768_A-12.zip' saved [407727028/40772
     Archive: uncased_L-12_H-768_A-12.zip
        creating: uncased_L-12_H-768_A-12/
       inflating: uncased_L-12_H-768_A-12/bert_model.ckpt.meta
       inflating: uncased L-12 H-768 A-12/bert model.ckpt.data-00000-of-00001
       inflating: uncased_L-12_H-768_A-12/vocab.txt
       inflating: uncased L-12 H-768 A-12/bert model.ckpt.index
       inflating: uncased_L-12_H-768_A-12/bert_config.json
!pwd
     /content
     sample data uncased L-12 H-768 A-12 uncased L-12 H-768 A-12.zip
# Start the BERT server
bert_command='bert-serving-start -model_dir /content/uncased_L-12_H-768_A-12'
process=subprocess.Popen(bert command.split(),stdout=subprocess.PIPE)
!nohup bert-serving-start -model dir=./uncased L-12 H-768 A-12 > out.file 2>&1 &
%tensorflow_version 1.x
     TensorFlow 1.x selected.
import tensorflow
print(tensorflow.__version__)
```

!15

1.15.2

Created wheel for GPUtil: filename=GPUtil-1.4.0-cp36-none-any.whl size=7411 sha256=

```
from bert_serving.client import BertClient
bc = BertClient()
# Load the training dataset
df = pd.read_csv('./train.csv')
print(df.head())
       id label
                                                          tweet
              0 #fingerprint #Pregnancy Test https://goo.gl/h1...
              0 Finally a transparant silicon case ^^ Thanks t...
              0 We love this! Would you go? #talk #makememorie...
    3 4
              0 I'm wired I know I'm George I was made that wa...
              1 What amazing service! Apple won't even talk to...
# Create a list of punctuation marks
#Ref google
puncts = [',', '.', '"', ':', ')', '(', '-', '!', '?', '|', ';', "'", '$', '&', '/', '[', ']', ']
 '.', '_', '{(', '}', '@', '^', '®', '`', '<', '→', '°', '€', '™', '>', '♥', '←', '×', '§
 # Code to replace punctuations with whitespaces
def clean_text(x):
   x = str(x)
   for punct in puncts:
       if punct in x:
           x = x.replace(punct, ' ')
   return x
df.tweet=df.tweet.apply(lambda x:clean_text(x))
df.tweet=df.tweet.apply(lambda x:re.sub(r'http\S+','',x))
df.tweet=df.tweet.apply(lambda x:re.sub(r'@[\w]*','',x))
df.tweet=df.tweet.apply(lambda x:x.lower())
df.tweet=df.tweet.apply(lambda x:' '.join(x.split()))
df.head()
```

```
id label
                                                           tweet
                 0
      0
          1
                      fingerprint pregnancy test goo gl h1mfqv andro...
      1
          2
                 0
                         finally a transparant silicon case thanks to m...
# Compute embeddings for training tweets using Bert Client encode function
# The model returns 768-dimensional embeddings
tweets=df.tweet
tweet_list=[word for word in tweets]
embeddings=bc.encode(tweet_list)
     /usr/local/lib/python3.6/dist-packages/bert_serving/client/__init__.py:299: UserWarni
     here is what you can do:
     - disable the length-check by create a new "BertClient(check_length=False)" when you
     - or, start a new server with a larger "max_seq_len"
       '- or, start a new server with a larger "max_seq_len"' % self.length_limit)
print(embeddings.shape)
     (7920, 768)
# save bert_train_new for reuse as it would take a really long time for conversion
pickle_out=open('bert_train.pickle','wb')
pickle.dump(embeddings,pickle_out)
pickle_out.close()
#loading test dataset
df1=pd.read_csv('test.csv')
df1.head()
           id
                                                    tweet
               I hate the new #iphone upgrade. Won't let me d...
      1 7922
                  currently shitting my fucking pants. #apple #i...
      2 7923 I'd like to puts some CD-ROMS on my iPad, is t...
        7924
                     My ipod is officially dead. I lost all my pict...
      4 7925
                   Been fighting iTunes all night! I only want th...
df1.tweet=df1.tweet.apply(lambda x:clean_text(x))
df1.tweet=df1.tweet.apply(lambda x:re.sub(r'http\S+','',x))
df1.tweet=df1.tweet.apply(lambda x:re.sub(r'@[\w]^*','',x))
df1.tweet=df1.tweet.apply(lambda x:x.lower())
df1.tweet=df1.tweet.apply(lambda x:' '.join(x.split()))
# Compute embeddings for training tweets using Bert Client encode function
# The model returns 768-dimensional embeddings
test tweets=df1.tweet
test tweet list=[word for word in test tweets]
```

```
test embeddings=bc.encode(test tweet list)
      print(test_embeddings.shape)
                   (1953, 768)
      # save bert_train_new for reuse as it would take a really long time for conversion
      pickle_out1=open('bert_test.pickle','wb')
      pickle.dump(test_embeddings,pickle_out1)
      pickle_out1.close()
Spacy Word2Vec
      pd.set option('display.max colwidth',100)
      !python -m spacy download en_vectors_web_lg
                  Collecting en_vectors_web_lg==2.1.0
                       Downloading <a href="https://github.com/explosion/spacy-models/releases/download/en vectors">https://github.com/explosion/spacy-models/releases/download/en vectors</a>
                                                                                           661.8MB 1.1MB/s
                  Requirement already satisfied: spacy<3.0.0,>=2.1.0 in /usr/local/lib/python3.6/dist-r
                  Requirement already satisfied: numpy>=1.15.0 in /usr/local/lib/python3.6/dist-package
                  Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.6/dist-r
                  Requirement already satisfied: thinc==7.4.0 in /usr/local/lib/python3.6/dist-packages
                  Requirement already satisfied: catalogue<1.1.0,>=0.0.7 in /usr/local/lib/python3.6/di
                  Requirement already satisfied: wasabi<1.1.0,>=0.4.0 in /usr/local/lib/python3.6/dist-
                  Requirement already satisfied: requests<3.0.0,>=2.13.0 in /usr/local/lib/python3.6/di
                  Requirement already satisfied: srsly<1.1.0,>=1.0.2 in /usr/local/lib/python3.6/dist-r
                  Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist
                  Requirement already satisfied: plac<1.2.0,>=0.9.6 in /usr/local/lib/python3.6/dist-page 1.2.0,>=0.9.6 in /usr/local/lib/py
                  Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/python3.6/
                  Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.6/dist-r
                  Requirement already satisfied: setuptools in /usr/local/lib/python3.6/dist-packages (
                  Requirement already satisfied: blis<0.5.0,>=0.4.0 in /usr/local/lib/python3.6/dist-page 1.0.5.0,>=0.4.0 in /usr/local/lib/python3.6/dist-page 2.0.5.0,>=0.4.0 in /usr/local/lib/python3.6/dist-page 2.0.5.0 in /usr/local/lib/python3.6/dist-page 2.0.5 in /usr/local/lib/python3.6/dist-page 
                  Requirement already satisfied: importlib-metadata>=0.20; python version < "3.8" in /u
                  Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.6/dist-pac
                  Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/
                  Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.6/dist-packages
                  Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-pa
                  Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.6/dist-packages (1
                  Building wheels for collected packages: en-vectors-web-lg
                        Building wheel for en-vectors-web-lg (setup.py) ... done
                       Created wheel for en-vectors-web-lg: filename=en_vectors_web_lg-2.1.0-cp36-none-any
                       Stored in directory: /tmp/pip-ephem-wheel-cache-c0k7l_x1/wheels/ce/3e/83/59647d0b45
                  Successfully built en-vectors-web-lg
                  Installing collected packages: en-vectors-web-lg
                  Successfully installed en-vectors-web-lg-2.1.0
                  ✓ Download and installation successful
                  You can now load the model via spacy.load('en_vectors_web_lg')
```

```
Requirement already satisfied: spacy<3.0.0,>=2.1.0 in /usr/local/lib/python3.6/dist-r
                 Requirement already satisfied: numpy>=1.15.0 in /usr/local/lib/python3.6/dist-package
                 Requirement already satisfied: plac<1.2.0,>=0.9.6 in /usr/local/lib/python3.6/dist-page 1.2.0,>=0.9.6 in /usr/local/lib/py
                 Requirement already satisfied: thinc==7.4.0 in /usr/local/lib/python3.6/dist-packages
                 Requirement already satisfied: setuptools in /usr/local/lib/python3.6/dist-packages (
                 Requirement already satisfied: wasabi<1.1.0,>=0.4.0 in /usr/local/lib/python3.6/dist-
                 Requirement already satisfied: srsly<1.1.0,>=1.0.2 in /usr/local/lib/python3.6/dist-r
                 Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist
                 Requirement already satisfied: catalogue<1.1.0,>=0.0.7 in /usr/local/lib/python3.6/di
                 Requirement already satisfied: blis<0.5.0,>=0.4.0 in /usr/local/lib/python3.6/dist-page 1.0.0 in /usr/local/li
                 Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.6/dist-r
                 Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.6/dist-r
                 Requirement already satisfied: requests<3.0.0,>=2.13.0 in /usr/local/lib/python3.6/di
                 Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/python3.6/
                 Requirement already satisfied: importlib-metadata>=0.20; python_version < "3.8" in /u
                 Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-page 1.00 in /usr/local/lib/
                 Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.6/dist-pac
                 Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.6/dist-packages
                 Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/
                 Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.6/dist-packages (1
                 ✓ Download and installation successful
                 You can now load the model via spacy.load('en vectors web lg')
import spacy
# Load the largest english language vector collection from Spacy
nlp = spacy.load('en_vectors_web_lg')
df['label'].value_counts(normalize = True)
                 0
                                  0.744192
                 1
                                  0.255808
                 Name: label, dtype: float64
# Function to lemmatize the tokens to their basic forms to normalize the tweet text
# and focus on key words for the classification tasks
def lemmatization(texts):
             output = []
             for i in texts:
                            s = [token.lemma_ for token in nlp(i)]
                            output.append(' '.join(s))
             return output
%%time
df.tweet=lemmatization(df.tweet)
df.tweet=df.tweet.str.replace('-PRON-','')
                 CPU times: user 1.35 s, sys: 4.36 ms, total: 1.35 s
```

Wall time: 1.36 s

Requirement already satisfied: en vectors web lg==2.1.0 from https://github.com/explosition.org/

```
%%time
```

```
df1.tweet=lemmatization(df1.tweet)
df1.tweet=df1.tweet.str.replace('-PRON-','')
```

CPU times: user 318 ms, sys: 84 μs, total: 318 ms

Wall time: 318 ms

nlp('having').vector

```
1.3405e-01, 3.7844e-04, 5.4900e-02, -2.5173e-01, -3.5485e-01,
-3.7260e-01, -1.7240e-03, 1.1956e+00, -4.1293e-01, 3.5877e-01,
5.1265e-03, -2.9626e-01, -2.4748e-01, 1.6286e-01, 7.5768e-02,
1.3535e-02, -7.9647e-02, -4.9073e-01, -1.0783e-01, -6.3812e-02,
-1.3171e-01, 1.8626e-01, 2.4554e-01, 2.5685e-01, 3.0148e-01,
-4.8167e-01, 4.0285e-01, -4.7838e-02, -8.0964e-02, -5.6645e-01,
 2.1666e-01, 1.1220e-01, 1.5485e-02, 3.1444e-01, -4.7426e-01,
3.0210e-01, -3.6470e-01, -3.4347e-01, 9.9283e-02, -8.5861e-02,
-8.2277e-02, -2.5866e-02, -4.7161e-02, -2.1301e-01, 2.6880e-01,
1.8113e-01, -2.0620e-01, -2.4319e-02, -1.5963e-01, 8.6472e-02,
1.8116e-01, 1.2205e-01, -4.6879e-01, 2.7622e-01, 3.7899e-02,
4.4370e-03, 2.6413e-01, 2.2721e-01, -1.7805e-02, 2.7563e-01,
2.6386e-01, 1.7431e-01, -6.1444e-02, -2.1381e-02, -2.7438e-02,
1.6243e-01, 2.8102e-01, -3.5839e-02, 1.7901e-01, 3.2328e-01,
-1.1513e-01, 1.3440e-01, -1.8181e-01, -5.0755e-01, 1.9801e-02,
-3.0611e-01, 2.8132e-01, -3.1478e-02, 1.8076e-01, -8.5850e-03,
-1.4519e-02, 2.8539e-02, -2.0772e-01, 1.8872e-01, -3.3428e-02,
2.5718e-01, 4.4756e-01, -2.1874e-01, 4.7900e-02, 1.4013e-01,
-3.2908e-01, 3.1017e-02, -1.5771e-02, -2.7796e-01, 2.0601e-01,
-1.3484e+00, -1.3698e-01, -8.7260e-02, -6.8283e-02, 1.7768e-01,
1.2368e-01, 2.2966e-01, 3.7684e-03, 3.6778e-02, -1.9610e-01,
-4.0696e-01, -1.2112e-01, 2.8510e-01, -2.4706e-01, 4.0122e-01,
2.9606e-01, 1.7297e-01, 5.7350e-01, 6.2956e-02, 3.7901e-01,
-1.9420e-02, -1.4721e-01, -3.1434e-01, -2.4116e-01, -2.2703e-01,
-1.5893e-03, 1.8312e-01, -3.2423e-01,
                                      1.5497e-01, 3.3933e-01,
-2.3480e-01, 1.5851e-02, 2.7963e-01, 1.8745e-02, -1.5975e-01,
-1.5019e+00, 5.0632e-02, 3.6933e-02, 1.0450e-01, -7.0496e-02,
-2.0645e-01, -7.0083e-02, -8.1474e-02, 1.8476e-01, -9.9499e-02,
-3.0478e-01, 7.6468e-02, -2.3014e-01, -7.0870e-02, -7.0931e-02,
8.1447e-02, 8.0975e-02, 4.3891e-01, 1.9877e-01, -3.2176e-01,
2.1967e-01, -5.7821e-01, 3.0394e-01, -1.2663e-01, -1.0427e-01,
-2.4780e-01, 2.6204e-01, 6.2570e-02, 9.1614e-02, 1.8825e-02,
-2.6012e-01, -4.1146e-01, 2.7580e-01, -4.9186e-03, -8.3340e-02,
-1.1895e-01, -3.8721e-01, 4.7886e-02, 9.9593e-02, -2.6970e-01,
-7.3007e-03, -3.7161e-01, -6.0079e-01, 4.3112e-02, -1.7589e-01,
-3.2411e-01, -2.6899e-01, 7.3743e-01, -1.7653e-01, -1.7557e-01,
1.6940e-01, 1.9966e-02, -1.3267e-01, 5.9843e-01, 2.3689e-01,
1.1431e-02, -8.2624e-02, 2.5213e-01, -5.1019e-01, 1.7412e-01,
4.0625e-01, -1.0041e-03, 2.7558e-01, 7.2856e-03, 3.6192e-01,
9.3313e-02, -4.0080e-01, -2.0661e-01, -5.1045e-03, 1.5150e-01,
-2.6760e-01, 2.6065e-01, -3.8441e-01, -4.5888e-02, -3.4107e-01,
 2.3661e-01, -2.5816e-01, -1.6351e-01, 1.4184e-01, 1.7698e-01,
-1.1873e-01, -7.8805e-02, -2.2065e-01, 2.1354e-01, 8.3310e-02,
-1.3151e-02, 1.6681e-01, 6.7123e-02, -1.4861e-01, -7.7549e-02,
1.3314e-01, -2.5016e-01, 3.0317e-02, -4.2529e-02, 2.6820e-01,
-2.5129e-01, 1.7177e-01, 8.6223e-02, -1.0212e-01, 1.1251e-01,
-6.6374e-02, 3.7500e-02, 2.6159e-01, 6.3398e-01, -7.4445e-02,
-2.2132e-03, -3.4139e-02, 1.3005e-01, -3.4528e-01, 2.7955e-02,
 1.4248e-01, -2.3346e-01, 3.2881e-01, 1.5303e-01, 1.7503e-01,
```

```
3.1437e-01, 4.3509e-01, -3.9043e-01, 5.4367e-01, 2.8549e-01,
             7.8270e-01, -2.2442e-02, 1.1466e-01, 5.1672e-01, -2.9182e-01,
            -4.3049e-02, -7.7364e-02, -3.9407e-01, -2.5879e-01, -3.4362e-01,
             2.9721e-01, -2.6811e-01, 8.9689e-02, 1.2101e-01, 5.0895e-01,
             2.8325e-01, 4.3377e-01, 9.8544e-02, 5.9706e-02, -1.3283e-02,
            -1.0903e-01, 2.1455e-01, -2.9188e-01, 1.6256e-01, 2.1777e-01,
            -1.4039e-01, -8.1819e-03, -3.7918e-01, -2.1583e-01, -1.8292e-01,
            -5.0702e-02, -7.3112e-02, -1.6639e-03, -1.7232e-02, 3.5350e-02],
           dtype=float32)
# Convert cleaned tweets into Spacy word vectors
# The model returns 300-dimensional embeddings
tweets=df.tweet
tweet_list=[nlp(word).vector for word in tweets]
X_tr=np.array(tweet_list)
test_tweets = df1.tweet
test_word_vec = [nlp(word).vector for word in test_tweets]
X_te = np.array(test_word_vec)
print(X_tr.shape, X_te.shape)
     (7920, 300) (1953, 300)
# Save Spacy_train_new
pickle_out = open("Spacy_train.pickle","wb")
pickle.dump(X_tr, pickle_out)
pickle_out.close()
# Save Spacy_test_new
pickle_out = open("Spacy_test.pickle","wb")
pickle.dump(X_te, pickle_out)
pickle out.close()
# Additional 'Optional' step for text normalization
# Import spaCy's language model
nlp1 = spacy.load('en', disable=['parser', 'ner'])
%tensorflow_version 1.x
     TensorFlow 1.x selected.
!pip install tensorflow-hub
     Requirement already satisfied: tensorflow-hub in /usr/local/lib/python3.6/dist-packas
     Requirement already satisfied: protobuf>=3.8.0 in /usr/local/lib/python3.6/dist-packa
     Requirement already satisfied: numpy>=1.12.0 in /usr/local/lib/python3.6/dist-package
     Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.6/dist-packages
     Requirement already satisfied: setuptools in /usr/local/lib/python3.6/dist-packages (
                                                                                        •
```

1.3949e-01, -5.0988e-02, -9.5092e-02, -5.1364e-02, 2.5831e-01,

```
!pip install tensorflow gpu==1.5.0
         Collecting tensorflow gpu==1.5.0
             Downloading <a href="https://files.pythonhosted.org/packages/d5/8b/094add4d2d667ddfef8672856">https://files.pythonhosted.org/packages/d5/8b/094add4d2d667ddfef8672856</a>
                                               201.9MB 85kB/s
         Requirement already satisfied: protobuf>=3.4.0 in /usr/local/lib/python3.6/dist-packa
         Requirement already satisfied: six>=1.10.0 in /usr/local/lib/python3.6/dist-packages
         Requirement already satisfied: wheel>=0.26 in /usr/local/lib/python3.6/dist-packages
         Requirement already satisfied: absl-py>=0.1.6 in /usr/local/lib/python3.6/dist-packas
         Collecting tensorflow-tensorboard<1.6.0,>=1.5.0
             Downloading <a href="https://files.pythonhosted.org/packages/cc/fa/91c06952517b4f1bc075545b6">https://files.pythonhosted.org/packages/cc/fa/91c06952517b4f1bc075545b6</a>
                                                                  3.0MB 50.0MB/s
         Requirement already satisfied: numpy>=1.12.1 in /usr/local/lib/python3.6/dist-package
         Requirement already satisfied: setuptools in /usr/local/lib/python3.6/dist-packages (
         Requirement already satisfied: werkzeug>=0.11.10 in /usr/local/lib/python3.6/dist-pac
         Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.6/dist-packation already satisfied: markdown already satisf
         Requirement already satisfied: bleach==1.5.0 in /usr/local/lib/python3.6/dist-package
         Requirement already satisfied: html5lib==0.9999999 in /usr/local/lib/python3.6/dist-r
         Requirement already satisfied: importlib-metadata; python_version < "3.8" in /usr/loc
         Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.6/dist-packages (1
         Installing collected packages: tensorflow-tensorboard, tensorflow-gpu
             Found existing installation: tensorflow-gpu 1.2.0
                 Uninstalling tensorflow-gpu-1.2.0:
                     Successfully uninstalled tensorflow-gpu-1.2.0
         Successfully installed tensorflow-gpu-1.5.0 tensorflow-tensorboard-1.5.1
import tensorflow hub as hub
import tensorflow as tf
elmo = hub.Module("https://tfhub.dev/google/elmo/2", trainable=True)
tf.test.gpu device name()
          '/device:GPU:0'
def elmo convert(x):
    embeddings=elmo(x.tolist(),signature='default',as dict=True)['elmo']
   with tf.Session() as sess:
       sess.run(tf.global_variables_initializer())
       sess.run(tf.tables initializer())
       # return average of ELMo features
       return sess.run(tf.reduce_mean(embeddings,1))
# Creating batches of 100 tweets to feed into elmo model at a time as it consumes high com
list train = [df[i:i+100]] for i in range(0,df.shape[0],100)]
list_test = [df1[i:i+100] for i in range(0,df1.shape[0],100)]
# Extract ELMo embeddings
elmo_train=[elmo_convert(x['tweet']) for x in list_train]
```

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```

```
# Concatenating converted batches into single array of train and test dataset embeddings
elmo_train_new=np.concatenate(elmo_train,axis=0)
elmo_test_new=np.concatenate(elmo_test,axis=0)
# save elmo_train
pickle_out = open("elmo_train.pickle","wb")
pickle.dump(elmo_train_new, pickle_out)
pickle_out.close()
# save elmo_test
pickle_out = open("elmo_test.pickle","wb")
pickle.dump(elmo_test_new, pickle_out)
pickle_out.close()
df_train=pd.read_csv('train.csv')
#adding tweet length
df_train['tweet_len_total']=df_train.tweet.str.len()
df_train.head()
         id label
                                                         tweet tweet_len_total
      0
         1
                0
```

```
#fingerprint #Pregnancy Test https://goo.gl/h1...
                                                                                    128
    2
                     Finally a transparant silicon case ^^ Thanks t...
1
            0
                                                                                   131
    3
2
            0 We love this! Would you go? #talk #makememorie...
                                                                                   123
3
    4
            0
                  I'm wired I know I'm George I was made that wa...
                                                                                    112
            1
4
    5
                   What amazing service! Apple won't even talk to...
                                                                                   124
```

```
#punctuation length as a feature
# Function to calculate the total length of punctuation marks in a tweet
def puncts_len(x):
    punct_list = []
    x = str(x)
    for punct in puncts:
        for char in x:
            if punct==char:
                punct_list.append(punct)
    return len(punct_list)

df_train['punc_len']=df_train.tweet.apply(lambda x:puncts_len(x))

df_train.head()
```

	id	label	tweet	tweet_len_total	punc_len
0	1	0	#fingerprint #Pregnancy Test https://goo.gl/h1	128	16
1	2	0	Finally a transparant silicon case ^^ Thanks t	131	17
2	3	0	We love this! Would you go? #talk #makememorie	123	18
_		_		440	

df_train.punc_len.describe()

count	7920.000000
mean	13.956692
std	8.406173
min	0.000000
25%	7.000000
50%	15.000000
75%	18.000000
max	59.000000

Name: punc_len, dtype: float64

df_train.info()

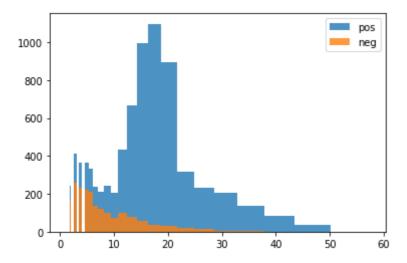
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7920 entries, 0 to 7919
Pata columns (total F columns):

Data columns (total 5 columns):

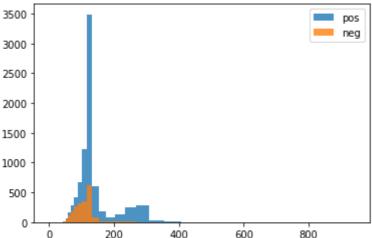
#	Column	Non-Null Count	Dtype
0	id	7920 non-null	int64
1	label	7920 non-null	int64
2	tweet	7920 non-null	object
3	<pre>tweet_len_total</pre>	7920 non-null	int64
4	punc_len	7920 non-null	int64

dtypes: int64(4), object(1)
memory usage: 309.5+ KB

```
bins=1.15**(np.arange(0,30))
plt.hist(df_train.punc_len,bins=bins,alpha=0.8)
plt.hist(df_train[df_train.label==1]['punc_len'],bins=bins,alpha=0.8)
plt.legend(('pos','neg'))
plt.show()
```



```
bins = 1.15**(np.arange(0,50))
plt.hist(df_train['tweet_len_total'],bins=bins,alpha=0.8)
plt.hist(df_train[df_train['label']==1]['tweet_len_total'],bins=bins,alpha=0.8)
plt.legend(('pos','neg'))
plt.show()
```



```
200
                            400
                                     600
                                             800
df_train.to_csv('more_features_train.csv')
# Similarly for test dataset loading new dataframe
df_test= pd.read_csv('test.csv')
#adding tweet length
df_test['tweet_len_total']=df_test.tweet.str.len()
df_test['punc_len']=df_test.tweet.apply(lambda x:puncts_len(x))
df_test.to_csv('more_features_test.csv')
# Load Spacy train Vectors
pickle_in = open("Spacy_train.pickle","rb")
spacy_train = pickle.load(pickle_in)
# Load Spacy_test Vectors
pickle_in = open("Spacy_test.pickle","rb")
spacy_test = pickle.load(pickle_in)
# Load BERT_train Vectors
pickle_in = open("bert_train.pickle","rb")
bert_train = pickle.load(pickle_in)
# Load BERT_test Vectors
pickle_in = open("bert_test.pickle","rb")
bert_test = pickle.load(pickle_in)
```

```
# Load ELMo_train Vectors
pickle_in = open("elmo_train.pickle","rb")
elmo_train = pickle.load(pickle_in)
# Load ELMo test Vectors
pickle_in = open("elmo_test.pickle","rb")
elmo_test = pickle.load(pickle_in)
# Create Spacy + BERT Vectors
sb_train=np.hstack((spacy_train,bert_train))
sb_test=np.hstack((spacy_test,bert_test))
# save spacy_bert_train
pickle_out = open("Spacy_bert_train.pickle","wb")
pickle.dump(sb_train, pickle_out)
pickle_out.close()
# save Spacy_bert_test
pickle_out = open("Spacy_bert_test.pickle","wb")
pickle.dump(sb_test, pickle_out)
pickle_out.close()
# Create BERT + ELMo Vectors
bert_elmo_train = np.hstack((bert_train, elmo_train))
bert_elmo_test = np.hstack((bert_test, elmo_test))
print(bert_elmo_train.shape, bert_elmo_test.shape)
# save bert_elmo_train
pickle_out = open("bert_elmo_train.pickle","wb")
pickle.dump(bert_elmo_train, pickle_out)
pickle_out.close()
# save bert_elmo_test
pickle_out = open("bert_elmo_test.pickle","wb")
pickle.dump(bert_elmo_test, pickle_out)
pickle_out.close()
     (7920, 1792) (1953, 1792)
# Create Spacy + ELMo Vectors
spacy_elmo_train = np.hstack((spacy_train, elmo_train))
spacy_elmo_test = np.hstack((spacy_test, elmo_test))
print(spacy_elmo_train.shape, spacy_elmo_test.shape)
# save Spacy_elmo_train
pickle_out = open("Spacy_elmo_train.pickle","wb")
pickle.dump(spacv elmo train. pickle out)
```

```
pickle_out.close()
# save Spacy_elmo_test
pickle out = open("Spacy elmo test.pickle","wb")
pickle.dump(spacy_elmo_test, pickle_out)
pickle_out.close()
     (7920, 1324) (1953, 1324)
# Create Spacy + BERT + ELMo Vectors
spacy_bert_elmo_train = np.hstack((spacy_train, bert_train, elmo_train))
spacy_bert_elmo_test = np.hstack((spacy_test, bert_test, elmo_test))
print(spacy_bert_elmo_train.shape, spacy_bert_elmo_test.shape)
# save Spacy_bert_elmo_train
pickle_out = open("Spacy_bert_elmo_train.pickle","wb")
pickle.dump(spacy_bert_elmo_train, pickle_out)
pickle_out.close()
# save Spacy_bert_elmo_test
pickle_out = open("Spacy_bert_elmo_test.pickle","wb")
pickle.dump(spacy_bert_elmo_test, pickle_out)
pickle_out.close()
     (7920, 2092) (1953, 2092)
# Load any variation of word embeddings from Spacy, BERT and ELMo and assign it to X varia
pickle_in = open("Spacy_bert_elmo_train.pickle","rb")
X=pickle.load(pickle_in)
X.shape
     (7920, 2092)
# Load the training dataset into a dataframe
df = pd.read_csv('more_features_train.csv')
print(df.head())
        Unnamed: 0 id ... tweet_len_total punc_len
     0
                0
                   1 ...
                                         128
                1 2 ...
     1
                                        131
                                                   17
     2
                2 3 ...
                                        123
                                                  18
     3
                 3 4 ...
                                                  17
                                        112
                 4
                                         124
                                                   5
     [5 rows x 6 columns]
```

df.label.value_counts()

```
0
          5894
     1
          2026
     Name: label, dtype: int64
y=df.label
from sklearn.preprocessing import MinMaxScaler
mms=MinMaxScaler(feature range=(-1,1))
tweet_len=mms.fit_transform(np.array(df.tweet_len_total).reshape(-1,1))
punc_len=mms.fit_transform(np.array(df.punc_len).reshape(-1,1))
final_X=np.hstack((X,tweet_len,punc_len))
print(final_X.shape)
     (7920, 2094)
# Split the training dataset into train and test subsets
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(final_X, y, test_size=0.1, random_stat
#we can apply any classification model
from sklearn import svm
svc=svm.LinearSVC()
from sklearn.pipeline import Pipeline
text_clf=Pipeline([('clf',svc)])
text_clf.fit(X_train,y_train)
     /usr/local/lib/python3.6/dist-packages/sklearn/svm/_base.py:947: ConvergenceWarning:
       "the number of iterations.", ConvergenceWarning)
     Pipeline(memory=None,
              steps=[('clf',
                      LinearSVC(C=1.0, class weight=None, dual=True,
                                fit_intercept=True, intercept_scaling=1,
                                loss='squared hinge', max iter=1000,
                                multi_class='ovr', penalty='12', random_state=None,
                                tol=0.0001, verbose=0))],
              verbose=False)
```

```
" Hake predictions
predictions = text clf.predict(X test)
print(predictions)
 00000010000100001000100000000010000001
 101000011000111001101010111011011000001
 0 1 0 0 0 0 0 0 1 0 0 1 0 0 0
from sklearn import metrics
metrics.confusion_matrix(y_test,predictions)
 array([[545, 38],
    [ 61, 148]])
metrics.classification_report(y_test,predictions)
                 support\n\n
                          0.90
      precision
          recall f1-score
                        0
 0.93
    0.92
        583\n
             1
                0.80
                   0.71
                      0.75
                          209\n\n
             0 88
                792\n
                        0.85
 accuracy
                   macro avø
metrics.accuracy_score(y_test,predictions)
 0.875
# Loading test dataset
df1 = pd.read csv('more features test.csv')
print(df1.head())
  Unnamed: 0
       id
        ... tweet len total
               punc len
      7921
 0
     0
              77
                 6
        . . .
 1
     1
      7922
                 13
             115
 2
     2
      7923
             104
                 9
        . . .
```

4

6

129

70

3

4

3

7924

7925

```
[5 rows x 5 columns]
# Dropping tweet column as it is no longer required for final submission leaving only the
df1 = df1.drop(['tweet'],axis=1)
print(df1.head())
        Unnamed: 0    id tweet_len_total punc_len
                0 7921
     0
                                       77
                                                  6
     1
                1 7922
                                      115
                                                 13
                2 7923
     2
                                     104
                                                  9
                 3 7924
     3
                                      129
                                                  4
     4
                 4 7925
                                       70
                                                  6
# Loading corresponding test tweets embeddings as loaded for the training dataset
pickle_in = open("Spacy_bert_elmo_test.pickle","rb")
test_X = pickle.load(pickle_in)
print(test_X.shape)
     (1953, 2092)
# Preparing test dataset for predictions by adding text features to the tweet embeddings d
tweet_len_arr_test = np.array(df1['tweet_len_total'])
tweet_punct_arr_test = np.array(df1['punc_len'])
print(tweet_len_arr_test.shape, tweet_punct_arr_test.shape)
tweet_len_norm_test = mms.fit_transform(tweet_len_arr_test.reshape(-1, 1))
tweet_punct_norm_test = mms.fit_transform(tweet_punct_arr_test.reshape(-1, 1))
print(tweet_len_norm_test.shape, tweet_punct_norm_test.shape)
test_X = np.hstack((test_X, tweet_len_norm_test, tweet_punct_norm_test))
print(test_X.shape)
     (1953,) (1953,)
     (1953, 1) (1953, 1)
     (1953, 2094)
# Making predictions using trained model for the test dataset for final submission
test_predictions = text_clf.predict(test_X)
print(test predictions)
     [1 \ 1 \ 0 \ \dots \ 0 \ 1 \ 0]
df1.columns
     Index(['Unnamed: 0', 'id', 'tweet_len_total', 'punc_len', 'label'], dtype='object')
# Adding predicted labels to the test dataframe
df1['label'] = test predictions
df1.drop(['Unnamed: 0','tweet_len_total', 'punc_len'],axis=1,inplace=True)
print(df1.head())
```

	id	label
0	7921	1
1	7922	1
2	7923	0
3	7924	1
4	7925	1

Saving the final predicted submission file to csv
df1.to_csv('ALL_SVM.csv', index=False)