# Sentiment Analysis on Twitter for Positive and negative tweets

#### **Python Code**

#### **Importing libraries**

- import pandas as pd
- import matplotlib.pyplot as plt
- from sklearn.model selection import train test split
- from sklearn.preprocessing import LabelEncoder
- from sklearn.metrics import confusion\_matrix, classification\_report, accuracy\_score
- from sklearn.manifold import TSNE
- from sklearn.feature extraction.text import TfidfVectorizer
- from keras.preprocessing.text import Tokenizer
- from keras.preprocessing.sequence import pad\_sequences
- from keras.models import Sequential
- from keras.layers import Activation, Dense, Dropout, Embedding, Flatten, Conv1D, MaxPooling1D, LSTM
- from keras import utils
- from keras.callbacks import ReduceLROnPlateau, EarlyStopping
- # nltk
- import nltk
- from nltk.corpus import stopwords
- from nltk.stem import SnowballStemmer
- # Word2vec
- import gensim
- # Utility
- import re
- import numpy as np
- import os
- from collections import Counter
- import logging
- import time
- import pickle
- import itertools
- # Set log
- logging.basicConfig(format='%(asctime)s: %(levelname)s: %(message)s', level=logging.INFO)
- nltk.download('stopwords')

#### Data Sets:

```
1.468E+09 Mon Apr 06 NO_QUERY Enespecia @switchfoot http://twitpic.com/2y1zl - Awww, that's a bummer. You should a got David Carr of Third Day to do it.; D
1.468E+09 Mon Apr 06 NO_QUERY scotthamilt is upset that he can't update his Facebook by texting it... and might cry as a result School today also. Blah!
1.468E+09 Mon Apr 06 NO_QUERY Mattycus
2.1468E+09 Mon Apr 06 NO_QUERY Karoli
2.1468E+09 Mon Apr 06 NO_QUERY Karoli
3.1468E+09 Mon Apr 06 NO_QUERY Karoli
4.468E+09 Mon Apr 06 NO_QUERY with the can't update his Facebook by texting it... and might cry as a result School today also. Blah!
3.1468E+09 Mon Apr 06 NO_QUERY with the can't update his Facebook by texting it... and might cry as a result School today also. Blah!
3.1468E+09 Mon Apr 06 NO_QUERY with the can't update his Facebook by texting it... and might cry as a result School today also. Blah!
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3.1468E+09 Mon Apr 06 NO_QUERY with the can't update his Facebook by texting it... and might cry as a result School today also. Blah!
3.1468E+09 Mon Apr 06 NO_QUERY cozz
4.1468E+09 Mon Apr 06 NO_QUERY prints he can't update his Facebook by texting it... and in the can't update his Facebook by texting it... and in the can't update his Facebook by texting it... and in the can't update his Facebook by the can't see you all over there.
4.1468E+09 Mon Apr 06 NO_QUERY prints he prints he can't update his Facebook by the can't see you all over the can't see you all over there.
5.1468E+09 Mon Apr 06 NO_QUERY bardonlauri just re-pierced my ears
6.1468E+09 Mon Apr 06 NO_QUERY bardonlauri just re-pierced my ears
6.1468E+09 Mon Apr 06 NO_QUERY bardonlauri just re-pierced my ears
6.1468E+09 Mon Apr 06 NO_QUERY bardonlauri just re-pierced my ears
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6.1468E+09 Mon Apr 06 NO_QUERY bardonlauri just re-pierced my ears
6.1
```

- DATASET COLUMNS = ["target", "ids", "date", "flag", "user", "text"]
- DATASET ENCODING = "ISO-8859-1"
- TRAIN SIZE = 0.8
- # TEXT CLENAING
- TEXT CLEANING RE = "(a)S+|https?:\S+|http?:\S|[^A-Za-z0-9]+"
- # WORD2VEC
- W2V SIZE = 300
- W2V WINDOW = 7
- W2V EPOCH = 32
- W2V MIN COUNT = 10
- # KERAS
- SEQUENCE LENGTH = 300
- EPOCHS = 8
- BATCH SIZE = 1024
- # SENTIMENT
- POSITIVE = "POSITIVE"
- NEGATIVE = "NEGATIVE"
- NEUTRAL = "NEUTRAL"
- SENTIMENT\_THRESHOLDS = (0.4, 0.7)
- # EXPORT
- KERAS MODEL = "model.h5"
- WORD2VEC\_MODEL = "model.w2v"
- TOKENIZER\_MODEL = "tokenizer.pkl"
- ENCODER\_MODEL = "encoder.pkl"
- #path ="/"
- #dataset\_filename = os.listdir(path)[0]
- #dataset\_path = os.path.join("..",path,dataset\_filename)
- #print("Open file:", dataset\_path)
- #df = pd.read\_csv(dataset\_path, encoding =DATASET\_ENCODING, names=DATASET\_COLUMNS)
- df = pd.read\_csv("D:/DataEntry/sdataset.csv", encoding =DATASET\_ENCODING, names=DATASET\_COLUMNS)

• print("Dataset size:", len(df))

#### **Machine Learning Models:**

#### 1-Sentiment Analysis

- def decode sentiment(label):
- return decode map[int(label)]
- df.target = df.target.apply(lambda x: decode sentiment(x))
- target cnt = Counter(df.target)
- plt.figure(figsize=(16,8))
- plt.bar(target cnt.keys(), target cnt.values())
- plt.title("Dataset labels distribuition")
- stop words = stopwords.words("english")
- stemmer = SnowballStemmer("english")
- def preprocess(text, stem=False):
- # Remove link, user and special characters
- text = re.sub(TEXT\_CLEANING\_RE, '', str(text).lower()).strip()
- tokens = []
- for token in text.split():
- if token not in stop\_words:
- if stem:
- tokens.append(stemmer.stem(token))
- else:
- tokens.append(token)
- return " ".join(tokens)
- df.text = df.text.apply(lambda x: preprocess(x))
- df\_train, df\_test = train\_test\_split(df, test\_size=1-TRAIN\_SIZE, random\_state=42)
- documents = [\_text.split() for \_text in df\_train.text]
- w2v\_model = gensim.models.word2vec.Word2Vec(size=W2V\_SIZE,
- window=W2V WINDOW,
- min\_count=W2V\_MIN\_COUNT,
- workers=8)
- w2v\_model.build\_vocab(documents)
- words = w2v model.wv.vocab.keys()
- vocab\_size = len(words)
- w2v model.train(documents, total examples=len(documents), epochs=W2V EPOCH)
- w2v model.most similar("love")
- tokenizer = Tokenizer()
- tokenizer.fit\_on\_texts(df\_train.text)
- vocab size = len(tokenizer.word index) + 1
- x\_train = pad\_sequences(tokenizer.texts\_to\_sequences(df\_train.text), maxlen=SEQUENCE\_LENGTH)

- x test = pad\_sequences(tokenizer.texts\_to\_sequences(df\_test.text), maxlen=SEQUENCE\_LENGTH)
- labels = df train.target.unique().tolist()
- labels.append(NEUTRAL)
- labels
- encoder = LabelEncoder()
- encoder.fit(df train.target.tolist())
- y\_train = encoder.transform(df\_train.target.tolist())
- y test = encoder.transform(df test.target.tolist())
- y\_train = y\_train.reshape(-1,1)
- $y_{\text{test}} = y_{\text{test.reshape}}(-1,1)$
- y\_train[:10]
- embedding matrix = np.zeros((vocab size, W2V SIZE))
- for word, i in tokenizer.word index.items():
- if word in w2v\_model.wv:
- embedding matrix[i] = w2v model.wv[word]
- print(embedding matrix.shape)
- embedding\_layer = Embedding(vocab\_size, W2V\_SIZE, weights=[embedding\_matrix], input\_length=SEQUENCE\_LENGTH, trainable=False)
- model = Sequential()
- model.add(embedding\_layer)
- model.add(Dropout(0.5))
- model.add(LSTM(100, dropout=0.2, recurrent dropout=0.2))
- model.add(Dense(1, activation='sigmoid'))
- model.summary()
- model.compile(loss='binary\_crossentropy',
- optimizer="adam",
- metrics=['accuracy'])
- callbacks = [ ReduceLROnPlateau(monitor='val\_loss', patience=5, cooldown=0),
- EarlyStopping(monitor='val acc', min delta=1e-4, patience=5)]
- history = model.fit(x\_train, y\_train,
- batch\_size=BATCH\_SIZE,
- epochs=EPOCHS,
- validation split=0.1,
- verbose=1,
- callbacks=callbacks)
- score = model.evaluate(x test, y test, batch size=BATCH SIZE)
- print()
- print("ACCURACY:",score[1])
- print("LOSS:",score[0])
- def decode sentiment(score, include neutral=True):
- if include neutral:
- label = NEUTRAL
- if score <= SENTIMENT THRESHOLDS[0]:
- label = NEGATIVE
- elif score >= SENTIMENT THRESHOLDS[1]:
- label = POSITIVE

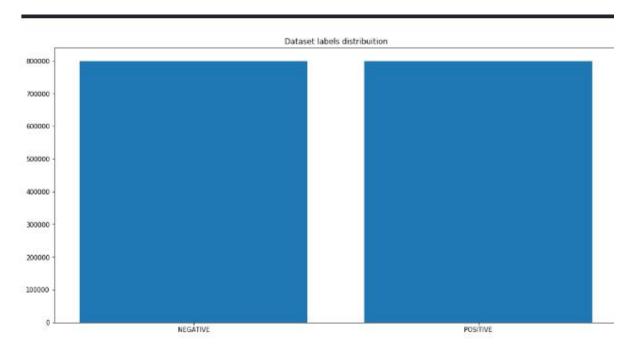
- return label
- else
- return NEGATIVE if score < 0.5 else POSITIVE
- def predict(text, include neutral=True):
- start at = time.time()
- # Tokenize text
- x\_test = pad\_sequences(tokenizer.texts\_to\_sequences([text]), maxlen=SEQUENCE\_LENGTH)
- # Predict
- score = model.predict([x\_test])[0]
- # Decode sentiment
- label = decode\_sentiment(score, include\_neutral=include\_neutral)
- return {"label": label, "score": float(score),
- "elapsed time": time.time()-start at}
- #Result =predict("I love the music")
- Result =predict("I hate the rain")
- print(Result)
- y\_pred\_1d = []
- y test 1d = list(df test.target)
- scores = model.predict(x test, verbose=1, batch size=8000)
- y\_pred\_1d = [decode\_sentiment(score, include\_neutral=False) for score in scores]
- accuracy score(y test 1d, y pred 1d)
- model.save(KERAS MODEL)
- w2v\_model.save(WORD2VEC\_MODEL)

#### Naïve Bayes

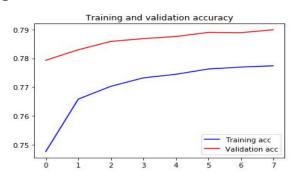
- print("naive bayes")
- from sklearn.naive\_bayes import GaussianNB
- model = GaussianNB()
- DATASET ENCODING = "ISO-8859-1"
- XDATASET COLUMNS = [ "text", "Results"]
- dt = pd.read\_csv("D:/DataEntry/Sampledata.csv", encoding =DATASET\_ENCODING, names=XDATASET\_COLUMNS)
- print(dt)
- X = dt.drop("Results", axis=1)
- y= dt["Results"]
- labelencoder = LabelEncoder()
- X["text"] = labelencoder.fit\_transform(X["text"])
- X\_train, X\_test, Y\_train, Y\_test = train\_test\_split(X,y,test\_size=0.25,random\_state=42)
- print(X test)
- model = GaussianNB()
- model.fit(X train, Y train)
- y pred = model.predict(X test)
- df1 = pd.concat([X\_test.reset\_index(drop='Tru e'),y\_pred.reset\_index(drop='Tru e')],axis=1)
- # Train the model
- #pickle.dump(tokenizer, open(TOKENIZER\_MODEL, "wb"), protocol=0)
- #pickle.dump(encoder, open(ENCODER\_MODEL, "wb"), protocol=0)

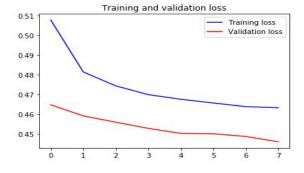
### **Visualizations:**

#### Bar Chart:

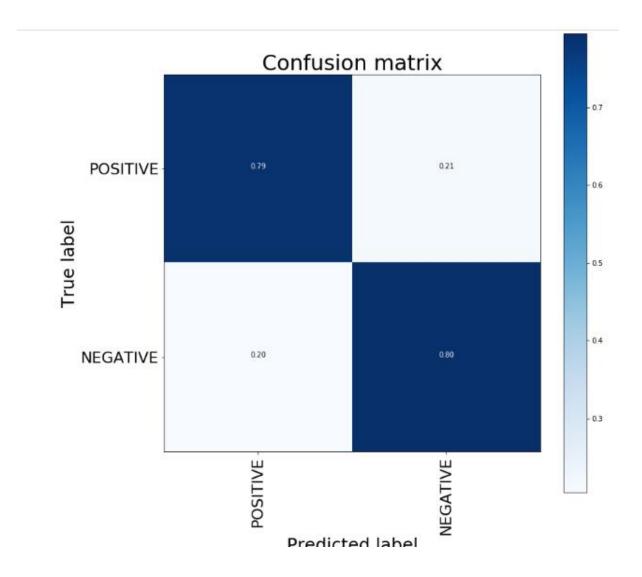


## Epochs:





# confusion\_matrix



Output:

Predict:

Result =predict("I love the music")

### Result:

Key	Туре	Size	Value
elapsed_time	float	1	0.3728067874908447
label	str	1	NEUTRAL
score	float	1	0.43893977999687195