

SPEECH SENTIMENT ANALYSIS

USING RECURSIVE NEURAL NETWORKS (LSTM)





- To achieve accuracy in predicting the sentiment of the speech
- Attempt to extend text-based sentiment analysis to speech
- Incorporated word sequencing on training data, and LSTM model for fitting the same

TRAINING DATASET

Sentiment140 Dataset

*Courtesy of Stanford

1.6 Million

Real Extracted Tweets

6 VARIABLES

Sentiment, Tweet ID, Date, Query,

Username, Text of the tweet

2 SENTIMENTS

Positive, negative

^{*} Citation: Go, A., Bhayani, R. and Huang, L., 2009. Twitter sentiment classification using distant supervision. CS224N Project Report, Stanford, 1(2009), p.12.

Procedure













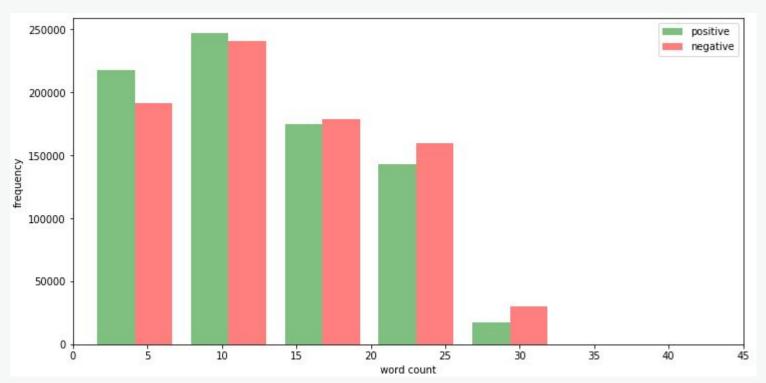




| | sentiment | id | date | flag | user | text | |
|--------------------------|-----------|------------|------------------------------|----------|-----------------|--|--|
| 0 | 0 | 1467810369 | Mon Apr 06 22:19:45 PDT 2009 | NO_QUERY | _TheSpecialOne_ | @switchfoot http://twitpic.com/2y1zl - Awww, t | |
| 1 | 0 | 1467810672 | Mon Apr 06 22:19:49 PDT 2009 | NO_QUERY | scotthamilton | is upset that he can't update his Facebook by | |
| 2 | 0 | 1467810917 | Mon Apr 06 22:19:53 PDT 2009 | NO_QUERY | mattycus | @Kenichan I dived many times for the ball. Man | |
| 3 | 0 | 1467811184 | Mon Apr 06 22:19:57 PDT 2009 | NO_QUERY | ElleCTF | my whole body feels itchy and like its on fire | |
| 4 | 0 | 1467811193 | Mon Apr 06 22:19:57 PDT 2009 | NO_QUERY | Karoli | @nationwideclass no, it's not behaving at all | |
| | *** | | | | | | |
| 1599995 | 4 | 2193601966 | Tue Jun 16 08:40:49 PDT 2009 | NO_QUERY | AmandaMarie1028 | Just woke up. Having no school is the best fee | |
| 1599996 | 4 | 2193601969 | Tue Jun 16 08:40:49 PDT 2009 | NO_QUERY | TheWDBoards | TheWDB.com - Very cool to hear old Walt interv | |
| 1599997 | 4 | 2193601991 | Tue Jun 16 08:40:49 PDT 2009 | NO_QUERY | bpbabe | Are you ready for your MoJo Makeover? Ask me f | |
| 1599998 | 4 | 2193602064 | Tue Jun 16 08:40:49 PDT 2009 | NO_QUERY | tinydiamondz | Happy 38th Birthday to my boo of allI time!!! | |
| 1599999 | 4 | 2193602129 | Tue Jun 16 08:40:50 PDT 2009 | NO_QUERY | RyanTrevMorris | happy #charitytuesday @theNSPCC @SparksCharity | |
| 1600000 rows × 6 columns | | | | | | | |



Wordcount v Freq by Sentiment





Pre-Processing



Clean tweets by removing link, user and special characters



Remove stopwords



Obtain tokenized versions of tweets and use stemming



Feature Engineering



Tweets' text converted to corpus

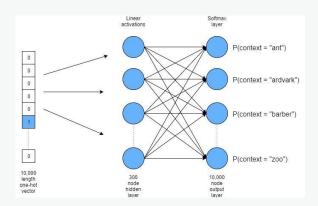


Corpus (text) converted to feature vector (numeric) using word2vec



Word feature vectors help ready input for the neural network





- Two layer neural net that processes text input
- Inputs text corpus and gives word feature vector as output
- Groups contextually similar words in the vector space,
 mathematically, using cosine distance

w2v model.most similar("silly")

```
[('dumb', 0.4478142559528351),
  ('stupid', 0.4135172963142395),
  ('funny', 0.40255165100097656),
  ('retarded', 0.3919326066970825),
  ('weird', 0.38109514117240906),
  ('wierd', 0.35878676176071167),
  ('childish', 0.35817572474479675),
  ('rude', 0.33793044090270996),
  ('mean', 0.3277094066143036),
  ('insensitive', 0.3249208927154541)]
```



Neural Network Model Definition



Neural RNN-LSTM Network defined

Model: "sequential 1" Layer (type) Output Shape Param # embedding 1 (Embedding) (None, 300, 300) 100652400 dropout 1 (Dropout) (None, 300, 300) 0 (None, 100) 160400 1stm 1 (LSTM) dense 1 (Dense) (None, 1) 101 Total params: 100,812,901 Trainable params: 160,501

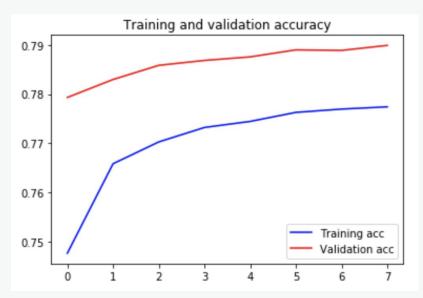
Non-trainable params: 100,652,400

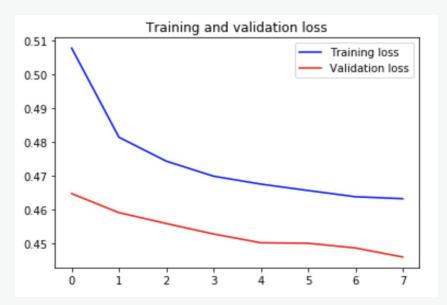


Fitting and Evaluation



Test and Validation Results







Fitting and Evaluation



Results on Test Set

• Accuracy: 0.8384401202201843

• Loss: 0.3755808174610138

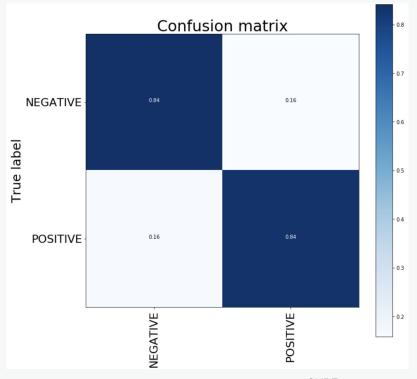


Fitting and Evaluation



Results on Test Set

| | precision | recall |
|----------|-----------|--------|
| NEGATIVE | 0.83 | 0.84 |
| POSITIVE | 0.84 | 0.84 |





Speech Sentiment Prediction



Inputting Speech from Microphone

- Speech recognition capabilities incorporated using the SpeechRecognition Python module
- Recognized speech is then converted to text for analysis



Speech Sentiment Prediction



Prediction

Classified sentiment label, as well as the sentiment score is returned

```
Go ahead and say something using the device microphone!

it's a beautiful day outside

predict(text)

{'label': 'POSITIVE', 'score': 0.9126543998718262}
```

Challenges



Detection of sarcasm, double-negation



Contextual-sentiment detection problem alleviated but still present



Computational time



Complexities in integrating tone and emotion detection



Difficulties with dialects and pace of speech



Business Applications

- Monitoring call center and customer support performance
- Gauging Consumer Responses
- Opinion Analysis via interviews
- Developing branding and marketing strategies
- Detection of hate speech in online videos and podcasts





- References -

- Sentiment140 dataset: Go, A., Bhayani, R. and Huang, L., 2009. Twitter sentiment classification using distant supervision. CS224N Project Report, Stanford, 1(2009), p.12.
- word2vec: Distributed Representations of Words and Phrases and their Compositionality, Tomas Mikolov, Ilya
 Sutskever, Kai Chen, Greg Corrado, Jeffrey Dean, Google Inc., (2013)
- SpeechRecognition: Zhang, Anthony. 2017. Speech Recognition (version 3.8).
- LSTM Neural Networks: Sepp Hochreite, Jürgen Schmidhuber, Neural Computation, Volume 9 Issue 8,
 November 15, 1997