

## **CSE 654 HW3 REPORT**

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# SENTIMENT ANALYSIS OF TURKISH MOVIE REVIEWS

### WITH WORD2VEC AND GAUSSIAN NAÏVE BAYES

#### **EXPLANATION**

The homework is about training a sentiment analysis with the help of Turkish sentences from Wikipedia and actual labeled movie reviews. We will use train a model with the help of unlabelled sentences(wikidump). This way we'll be using much more texts.

#### **COMBINING FILES**

Before training the data, preprocessing should be made. Also before giving our file to word2vec program we should be aware that our file doesn't contain weird symbols, newlines, tabs etc. I did this preprocessing in the file named file\_operations.py. It fixes the common errors like file format, decoding/encoding errors etc.

The input files are named as follows: "test.csv" which is our labelled test set, "train.scv" which is our labelled training movie data set and finally the "trwiki" file which contains Turkish Wikipedia reviews. In my folder I only included 5% of actual trwiki dump in order to upload to Moodle. The scripts combines these files then outputs a space separated words with the file named with "word2vecinput".

The combination of three files (word2vecinput file) in this form at the end:

film çok vasat hiçbir özelligi yok tavsiye etmem eski trk filmlerini izlememi o cok guzeldi filmin icinde hissettim kendimi sonu biraz daha degisik olsa super onemli bir kaniti hirschin genç yastaki ustaligi doga ve insanin lirik iliskisi yasama arzusu rahatlikla izlenebilir gereksiz bir film we gereksiz oyuncular dana çok fazla dövüs sahnesine ragmen çok sikici bir film izlemesinizde olur yan eyen arkadaslara tavsiyem sinemaya gidecek paraniz varsa paranizi bu filmle çöpe erinden biriydi oldukça farkli olmus çok begendim ben filme dram aksiyon denmi: ederim gerçekten bu oyunculara layik olmayan bir film çekilmis bir iki aksiyon tefen daha gercekci yorumlar yazin bu film bu kadar puanida kesinlikle hak etmiyo

FIGURE 1: CONTENT OF WORD2VEC INPUT FILE

#### **GENERATING VECTORS WITH WORD2VEC**

After creating the input file for word2vec program now we can create our vectors with it. I tried running word2vec program with different parameters but the default parameters with the vector size 100 gave the best results.

```
/Desktop/word2vec-master$ ./word2vec -train word2vecinput -output vectors -min-count = 1
Starting training using file word2vecinput
Vocab size: 64058
Words in train file: 562569
Alpha: 0.000106 Progress: 100.66% Words/thread/sec: 221.91k patoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/papatoglu@ubuntu:~/
```

FIGURE 2: CREATING VECTORS WITH WORD2VEC

No we've created the "word" vectors but what we need is sentence vectors. There are several ways to calculate the feature vector of a sentence. These are:

- 1. Calculating the average of word vectors.
- 2. Calculating the average of word vectors with their TF-IDF values.

These calculations will make an impact on probability calculation. In homework it says provide atleast 2 ways but I only implemented the first one. It simply tokenizes the sentence and according to word's vector score calculating the average of a sentence. We will use these comment vectors while training our model.

```
kötüyse 0.002642 -0.004564 -0.001119 -0.003384 -0.003334 -0.000005 0.0029 çekmenin 0.004631 -0.000352 -0.007019 0.003903 -0.015771 -0.022751 -0.004 izlemenisi 0.010329 -0.026544 0.022713 -0.015573 0.028809 -0.025592 -0.00 kalmistim -0.017033 -0.006224 0.000951 -0.012175 -0.005261 0.005666 0.026 özelllikle 0.010105 -0.032191 0.045078 -0.007272 0.008146 -0.028305 0.008 önemliside -0.022158 -0.006802 0.000828 -0.011347 -0.037074 -0.039191 0.0 çagdan 0.005318 -0.021060 -0.018921 0.012120 0.006827 0.024330 -0.012424 reklama 0.009523 -0.026586 -0.006013 -0.025535 0.066915 -0.004067 0.01207 edindi -0.004580 -0.007601 -0.005284 0.028625 0.022903 0.021653 -0.014801 askla 0.004146 -0.004634 -0.001582 -0.000442 0.002086 0.001770 -0.001313 aglat -0.022772 0.014816 0.017567 0.002160 -0.029847 0.008301 0.005022 -0.0005025 -0.0005025 -0.0005022 -0.0005025 -0.0005025 -0.0005022 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005025 -0.0005050 -0.0005025 -0.0005050 -0.0005050 -0.0005025 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.0005050 -0.00
```

FIGURE 3: THE OUTPUT FILE OF WORD2VEC PROGRAM

#### **CONTINIOUS NAÏVE BAYES AND TESTING MODEL**

Instead of using classical Naïve Bayes we will use Gaussian Naïve Bayes because our model is continuous. So we won't be doing counting as shown in the book.

**(b)** If  $A_k$  is continuous-valued, then we need to do a bit more work, but the calculation is pretty straightforward. A continuous-valued attribute is typically assumed to have a Gaussian distribution with a mean  $\mu$  and standard deviation  $\sigma$ , defined by

$$g(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}},$$
 (8.13)

FIGURE 4: GAUSSIAN NAIVE BAYES ON DAN JURAFSKY'S NLP BOOK

I couldn't implement this algorithm on my own so I used a library which does this calculation behind the scenes. A part of my code and the final result is like:

```
model = GaussianNB()
model.fit(np.array(sentence_vectors), np.array(labels))
prediction = model.predict(np.array(sentence_vectors_test))
print(f1_score(np.array(test_labels),prediction, average = 'micro'))
```

FIGURE 5: TESTING THE MODEL

Finally my accuracy was always between 0.65-0.67:

PS C:\Users\patog\One 0.6669167291822956

FIGURE 6: FINAL SCORE OF THE MODEL