Untitled1

October 26, 2019

1 Homework 2 Question 1: Word2vec - Word Embeddings

1.0.1 Github repo: https://github.com/anjaliverma96/avw4127_msia414_2019

```
import os
        os.chdir("20-newsgroups/")
        import glob
        read_files = glob.glob("*.txt")
        with open("final_text.txt", "wb") as outfile:
            for f in read_files:
                i+=1
                if(i==5):
                    break
                with open(f, "rb") as infile:
                    outfile.write(infile.read())
        #### READ IN THE FINAL TEXT FILE
        with open("final_text.txt",encoding="utf8", errors='ignore') as file:
            test_text = file.read()
1.1 Pre-Processing
In [3]: #### Create a list of all emails in the text by splitting on the word 'Newsgroup:' (Si
        email_list = test_text.split("Newsgroup:")
In [4]: #### Example of an email from the list
        email_list[5000]
Out[4]: " misc.forsale\nDocument_id: 75856\nFrom: mmm@cup.portal.com (Mark Robert Thorson)\nSu
```

In [2]: #### WRITE TEXT FILES FROM 5 SUB-FOLDERS THAT WERE PART OF THE NEWSGROUP FOLDER INTO O.

In [41]: #### Import necessary libraries

import re

```
import datetime
         import string
         import nltk
         from nltk.tokenize import sent_tokenize, word_tokenize, RegexpTokenizer
         from nltk.corpus import stopwords
         from nltk.stem.wordnet import WordNetLemmatizer
         from nltk.stem.porter import PorterStemmer
         from nltk.stem.wordnet import WordNetLemmatizer
         import numpy as np
In [6]: #### Download necessary packages
       nltk.download()
        nltk.download('stopwords')
        nltk.download('punkt')
        SENT_DETECTOR = nltk.data.load('tokenizers/punkt/english.pickle')
showing info https://raw.githubusercontent.com/nltk/nltk_data/gh-pages/index.xml
[nltk_data] Downloading package stopwords to
                /Users/anjaliverma/nltk_data...
[nltk_data]
             Package stopwords is already up-to-date!
[nltk_data]
[nltk_data] Downloading package punkt to
[nltk data]
                /Users/anjaliverma/nltk_data...
[nltk_data]
             Package punkt is already up-to-date!
In [7]: #### Define function that performs the following pre processing steps at once :
        def preprocessing_nltk(text):
           now = datetime.datetime.now()
            #### Create tokenizer that only tokenizes alpha-numeric words
            tokenizer = RegexpTokenizer(r'\w+')
            #### Convert text to lower case, tokenize text and remove numeric tokens
            revised_tokens = [word for word in tokenizer.tokenize(text.lower()) if word.isalpha
            #### Remove stopwords
            words = [w for w in revised_tokens if w not in stopwords.words('english')]
            #### Lemmatize tokens obtained after removing stopwords
            wnl = WordNetLemmatizer()
            tagged = nltk.pos_tag(words)
            lem_list = []
            for word, tag in tagged:
                wntag = tag[0].lower()
                wntag = wntag if wntag in ['a', 'r', 'n', 'v'] else None
```

```
if not wntag:
                    lemma = word
                else:
                    lemma = wnl.lemmatize(word, wntag)
                lem list.append(lemma)
            #lem_text = " ".join(lemma for lemma in lem_list)
            #print("Took %s"%(datetime.datetime.now()-now))
            return lem_list
In [8]: #### Example of a pre-processed email within the document
        doc = email_list[4000]
        print(preprocessing_nltk(doc))
['misc', 'forsale', 'subject', 'diamond', 'stealth', 'svga', 'sale', 'cleveland', 'freenet', '
In [9]: #### Loop through each email in the email list to preprocess each email
        #### Store each list within list processed_emails
       processed_emails = []
        for i in range(len(email_list)):
            processed_emails.append(preprocessing_nltk(email_list[i]))
In [10]: del processed_emails[0]
In [11]: print(processed_emails[500])
['sci', 'crypt', 'rschnapp', 'metaflow', 'com', 'rus', 'schnapp', 'subject', 'tap', 'code', 'g
In [12]: #### Write the preprocessed emails to a text file
         with open("anjali_verma_preprocessed_emails.txt", "w") as fobj:
             for x in processed_emails:
                 doc = " ".join(lemma for lemma in x)
                 fobj.write(doc + "\n")
1.2 Word2Vec: Creating word embeddings
In [13]: #### Example of the data in desired format for word embeddings
         #### The dataset is in the form of a list of list of tokens for each document (Newsgr
         print(processed_emails[500:502])
[['sci', 'crypt', 'rschnapp', 'metaflow', 'com', 'rus', 'schnapp', 'subject', 'tap', 'code', ';
In [14]: import gensim, logging
         from gensim.models import Word2Vec
         logging.basicConfig(format='%(asctime)s : %(levelname)s : %(message)s', level=logging
```

```
In [46]: def cosine_distance (model, word, target_list , num) :
               cosine_dict ={}
               word_list = []
               a = model[word]
               for item in target_list :
                    if item != word :
                         b = model [item]
                         cos_sim = np.dot(a, b)/(np.linalg.norm(a)*np.linalg.norm(b))
                         cosine_dict[item] = cos_sim
               dist_sort=sorted(cosine_dict.items(), key=lambda dist: dist[1],reverse = True) ##
               for item in dist_sort:
                    word_list.append((item[0], item[1]))
               return word_list[0:num]
In [35]: target_list= list(set([y for x in processed_emails for y in x ]))
In [37]: print(target_list[0:10])
['esp', 'spiffy', 'pixie', 'supertwist', 'extroverted', 'posteriorly', 'chow', 'repudiate', 'in
1.3 SKIP-GRAM MODEL (model parameter sg = 1)
In [66]: #### train word2vec
          model = Word2Vec(processed_emails, min_count=1,size= 50,workers=3, window =3, sg = 1)
2019-10-26 22:20:29,259: WARNING: consider setting layer size to a multiple of 4 for greater
2019-10-26 22:20:29,261 : INFO : collecting all words and their counts
2019-10-26 22:20:29,263 : INFO : PROGRESS: at sentence #0, processed 0 words, keeping 0 word t
2019-10-26 22:20:29,503 : INFO : collected 30557 word types from a corpus of 1122850 raw words
2019-10-26 22:20:29,503 : INFO : Loading a fresh vocabulary
2019-10-26 22:20:29,739 : INFO : effective_min_count=1 retains 30557 unique words (100% of original)
2019-10-26 22:20:29,740 : INFO : effective_min_count=1 leaves 1122850 word corpus (100% of original contents)
2019-10-26 22:20:29,840 : INFO : deleting the raw counts dictionary of 30557 items
2019-10-26 22:20:29,841 : INFO : sample=0.001 downsamples 22 most-common words
2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29,842 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:20:29)
2019-10-26 22:20:29,941 : INFO : estimated required memory for 30557 words and 50 dimensions:
2019-10-26 22:20:29,942 : INFO : resetting layer weights
2019-10-26 22:20:30,253 : INFO : training model with 3 workers on 30557 vocabulary and 50 feat
2019-10-26 22:20:31,269 : INFO : EPOCH 1 - PROGRESS: at 60.21% examples, 606820 words/s, in_qs
2019-10-26 22:20:31,990 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:20:32,002 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:20:32,012 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:20:32,013 : INFO : EPOCH - 1 : training on 1122850 raw words (1089767 effective
2019-10-26 22:20:33,030 : INFO : EPOCH 2 - PROGRESS: at 60.21% examples, 606468 words/s, in_qs
2019-10-26 22:20:33,744 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:20:33,747 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:20:33,754 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:20:33,755 : INFO : EPOCH - 2 : training on 1122850 raw words (1089567 effective
```

```
2019-10-26 22:20:34,780 : INFO : EPOCH 3 - PROGRESS: at 65.60% examples, 639189 words/s, in_qs
2019-10-26 22:20:35,455 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:20:35,469 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:20:35,493 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:20:35,494 : INFO : EPOCH - 3 : training on 1122850 raw words (1089755 effective
2019-10-26 22:20:36,514 : INFO : EPOCH 4 - PROGRESS: at 55.53% examples, 576437 words/s, in_qs
2019-10-26 22:20:37,283 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:20:37,297 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:20:37,309 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:20:37,310 : INFO : EPOCH - 4 : training on 1122850 raw words (1089753 effective
2019-10-26 22:20:38,321 : INFO : EPOCH 5 - PROGRESS: at 62.97% examples, 629001 words/s, in qs
2019-10-26 22:20:39,004 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:20:39,014 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:20:39,023 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:20:39,024 : INFO : EPOCH - 5 : training on 1122850 raw words (1089867 effective
2019-10-26 22:20:39,025 : INFO : training on a 5614250 raw words (5448709 effective words) too
In [67]: model.save("word2vec1.model")
        model = Word2Vec.load("word2vec1.model")
2019-10-26 22:20:43,677 : INFO : saving Word2Vec object under word2vec1.model, separately None
2019-10-26 22:20:43,678 : INFO : not storing attribute vectors_norm
2019-10-26 22:20:43,679 : INFO : not storing attribute cum_table
2019-10-26 22:20:43,886 : INFO : saved word2vec1.model
2019-10-26 22:20:43,887 : INFO : loading Word2Vec object from word2vec1.model
2019-10-26 22:20:44,031 : INFO : loading wv recursively from word2vec1.model.wv.* with mmap=No:
2019-10-26 22:20:44,031 : INFO : setting ignored attribute vectors_norm to None
2019-10-26 22:20:44,032 : INFO : loading vocabulary recursively from word2vec1.model.vocabulary
2019-10-26 22:20:44,033 : INFO : loading trainables recursively from word2vec1.model.trainable
2019-10-26 22:20:44,033 : INFO : setting ignored attribute cum_table to None
2019-10-26 22:20:44,035 : INFO : loaded word2vec1.model
In [49]: w1 = 'databases'
In [50]: model[w1]
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
Out[50]: array([-0.16538486, -0.22892961, -0.12038412, 0.39569497, 0.10224129,
                 0.11696152, 0.41309643, 0.15227576, -0.0053171, 0.329913,
                0.00425125, 0.14172778, -0.04164088, -0.01574311, 0.269511
                0.2694517 , 0.2946141 , -0.05229336 , 0.02670644 , -0.5384178 ,
```

0.2846486, 0.00451329, -0.22964086, 0.29623625, -0.12663527, -0.36935067, -0.03723044, -0.37925726, -0.34062475, -0.40986732, -0.26562893, -0.20815687, -0.15960026, -0.03676778, 0.05217892,

```
-0.0215673 , 0.33174634, 0.3337871 , 0.35224402, -0.2900794 ],
                                dtype=float32)
In [51]: model.similarity(w1, 'computer')
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
    """Entry point for launching an IPython kernel.
Out [51]: 0.5401499
In [52]: model.wv.most_similar(positive = w1,topn=5)
Out[52]: [('payware', 0.9192549586296082),
                      ('vital', 0.9152900576591492),
                      ('adminstrator', 0.9141732454299927),
                      ('uncompressor', 0.9132228493690491),
                      ('factoring', 0.9109092950820923)]
In [53]: cosine_distance (model, 'databases', target_list,5)
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
    after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWeets/
    import sys
Out[53]: [('payware', 0.919255),
                      ('vital', 0.91529),
                      ('adminstrator', 0.9141732),
                      ('uncompressor', 0.91322285),
                      ('factoring', 0.9109094)]
1.4 Changing model parameter window form 3 to 5
In [68]: model1 = Word2Vec(processed_emails, min_count=1, size= 50, workers=3, window =5, sg = 1
                   model1.save("word2vec2.model")
                   model1 = Word2Vec.load("word2vec2.model")
2019-10-26 22:21:01,935 : WARNING : consider setting layer size to a multiple of 4 for greater
2019-10-26 22:21:01,937 : INFO : collecting all words and their counts
2019-10-26 22:21:01,937 : INFO : PROGRESS: at sentence #0, processed 0 words, keeping 0 word to
2019-10-26 22:21:02,194 : INFO : collected 30557 word types from a corpus of 1122850 raw words
2019-10-26 22:21:02,195 : INFO : Loading a fresh vocabulary
2019-10-26 22:21:02,258 : INFO : effective_min_count=1 retains 30557 unique words (100% of original states of the count of
```

0.33948576, -0.3676621 , -0.09999277, 0.06539803, -0.21091425, 0.15911348, 0.21590903, 0.50502086, -0.06880483, -0.22672614,

2019-10-26 22:21:02,364: INFO: deleting the raw counts dictionary of 30557 items

2019-10-26 22:21:02,260 : INFO : effective_min_count=1 leaves 1122850 word corpus (100% of original contents)

```
2019-10-26 22:21:02,366 : INFO : sample=0.001 downsamples 22 most-common words
2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02,366 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:02)
2019-10-26 22:21:02,460 : INFO : estimated required memory for 30557 words and 50 dimensions:
2019-10-26 22:21:02,461 : INFO : resetting layer weights
2019-10-26 22:21:02,750 : INFO : training model with 3 workers on 30557 vocabulary and 50 feat
2019-10-26 22:21:03,775 : INFO : EPOCH 1 - PROGRESS: at 40.53% examples, 460818 words/s, in_qs
2019-10-26 22:21:04,802 : INFO : EPOCH 1 - PROGRESS: at 92.23% examples, 472590 words/s, in qs
2019-10-26 22:21:05,080 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:05,100 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:05,119 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:05,120 : INFO : EPOCH - 1 : training on 1122850 raw words (1089889 effective
2019-10-26 22:21:06,135 : INFO : EPOCH 2 - PROGRESS: at 28.47% examples, 373311 words/s, in qs
2019-10-26 22:21:07,148 : INFO : EPOCH 2 - PROGRESS: at 84.28% examples, 421828 words/s, in qs
2019-10-26 22:21:07,564 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:07,600 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:07,618 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:07,619 : INFO : EPOCH - 2 : training on 1122850 raw words (1089935 effective
2019-10-26 22:21:08,641 : INFO : EPOCH 3 - PROGRESS: at 32.67% examples, 405770 words/s, in qs
2019-10-26 22:21:09,649 : INFO : EPOCH 3 - PROGRESS: at 87.36% examples, 444913 words/s, in_qs
2019-10-26 22:21:10,003 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:10,015 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:10,035 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:10,036 : INFO : EPOCH - 3 : training on 1122850 raw words (1089538 effective
2019-10-26 22:21:11,051 : INFO : EPOCH 4 - PROGRESS: at 31.85% examples, 398562 words/s, in_qs
2019-10-26 22:21:12,058 : INFO : EPOCH 4 - PROGRESS: at 84.28% examples, 423006 words/s, in_qs
2019-10-26 22:21:12,541 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:12,575 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:12,583 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:12,584 : INFO : EPOCH - 4 : training on 1122850 raw words (1089916 effective
2019-10-26 22:21:13,597 : INFO : EPOCH 5 - PROGRESS: at 41.83% examples, 475363 words/s, in_qs
2019-10-26 22:21:14,610 : INFO : EPOCH 5 - PROGRESS: at 93.51% examples, 487974 words/s, in qs
2019-10-26 22:21:14,820 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:14,846 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:14,865 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:14,866 : INFO : EPOCH - 5 : training on 1122850 raw words (1089618 effective
2019-10-26 22:21:14,866 : INFO : training on a 5614250 raw words (5448896 effective words) too
2019-10-26 22:21:14,875 : INFO : saving Word2Vec object under word2vec2.model, separately None
2019-10-26 22:21:14,876 : INFO : not storing attribute vectors_norm
2019-10-26 22:21:14,877 : INFO : not storing attribute cum_table
2019-10-26 22:21:15,096 : INFO : saved word2vec2.model
2019-10-26 22:21:15,098 : INFO : loading Word2Vec object from word2vec2.model
2019-10-26 22:21:15,236: INFO: loading wv recursively from word2vec2.model.wv.* with mmap=No.
2019-10-26 22:21:15,236 : INFO : setting ignored attribute vectors_norm to None
2019-10-26 22:21:15,237 : INFO : loading vocabulary recursively from word2vec2.model.vocabular
2019-10-26 22:21:15,237 : INFO : loading trainables recursively from word2vec2.model.trainable
2019-10-26 22:21:15,238 : INFO : setting ignored attribute cum table to None
2019-10-26 22:21:15,239 : INFO : loaded word2vec2.model
```

Euclidean Similarity of word 'databases' to computer is: 0.5401499

import sys

Top 5 words similar to the given word according to euclidean similarity: [('payware', 0.91925

Top 5 words similar to the given word according to cosine similarity: [('payware', 0.919255),

1.5 CONTINUOUS BAG OF WORDS (model parameter sg = 0)

```
In [69]: #### train word2vec
                 new_model = Word2Vec(processed_emails, min_count=1,size=50, workers=3, window=3, sg =
2019-10-26 22:21:30,374: WARNING: consider setting layer size to a multiple of 4 for greater
2019-10-26 22:21:30,375 : INFO : collecting all words and their counts
2019-10-26 22:21:30,376 : INFO : PROGRESS: at sentence #0, processed 0 words, keeping 0 word to
2019-10-26 22:21:30,610 : INFO : collected 30557 word types from a corpus of 1122850 raw words
2019-10-26 22:21:30,610 : INFO : Loading a fresh vocabulary
2019-10-26 22:21:30,663 : INFO : effective_min_count=1 retains 30557 unique words (100% of original states)
2019-10-26 22:21:30,663 : INFO : effective_min_count=1 leaves 1122850 word corpus (100% of original contents)
2019-10-26 22:21:30,763 : INFO : deleting the raw counts dictionary of 30557 items
2019-10-26 22:21:30,765 : INFO : sample=0.001 downsamples 22 most-common words
2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated 1089714 word corpus (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves estimated (97.0% of page 2019-10-26 22:21:30,766 : INFO : downsampling leaves (97.0% of page 2019-10-26 22:21:30,766 : INFO : downs
2019-10-26 22:21:30,851 : INFO : estimated required memory for 30557 words and 50 dimensions:
2019-10-26 22:21:30,852 : INFO : resetting layer weights
2019-10-26 22:21:31,137 : INFO : training model with 3 workers on 30557 vocabulary and 50 feat
2019-10-26 22:21:31,881 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:31,888 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:31,890 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:31,890 : INFO : EPOCH - 1 : training on 1122850 raw words (1089837 effective
2019-10-26 22:21:32,692 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:32,701 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:32,702 : INFO : worker thread finished; awaiting finish of O more threads
2019-10-26 22:21:32,703 : INFO : EPOCH - 2 : training on 1122850 raw words (1089652 effective
2019-10-26 22:21:33,444 : INFO : worker thread finished; awaiting finish of 2 more threads
```

2019-10-26 22:21:33,452 : INFO : worker thread finished; awaiting finish of 1 more threads

```
2019-10-26 22:21:33,453 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:33,453 : INFO : EPOCH - 3 : training on 1122850 raw words (1089582 effective
2019-10-26 22:21:34,273 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:34,283 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:34,284 : INFO : worker thread finished; awaiting finish of O more threads
2019-10-26 22:21:34,285 : INFO : EPOCH - 4 : training on 1122850 raw words (1089859 effective
2019-10-26 22:21:35,047 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:21:35,056 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:21:35,057 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:21:35,058 : INFO : EPOCH - 5 : training on 1122850 raw words (1089919 effective
2019-10-26 22:21:35,059: INFO: training on a 5614250 raw words (5448849 effective words) too
In [70]: new_model.save("word2vec3.model")
        new_model = Word2Vec.load("word2vec3.model")
2019-10-26 22:21:44,765 : INFO : saving Word2Vec object under word2vec3.model, separately None
2019-10-26 22:21:44,766 : INFO : not storing attribute vectors_norm
2019-10-26 22:21:44,767 : INFO : not storing attribute cum_table
2019-10-26 22:21:44,980 : INFO : saved word2vec3.model
2019-10-26 22:21:44,981 : INFO : loading Word2Vec object from word2vec3.model
2019-10-26 22:21:45,343 : INFO : loading wv recursively from word2vec3.model.wv.* with mmap=Nor
2019-10-26 22:21:45,343 : INFO : setting ignored attribute vectors_norm to None
2019-10-26 22:21:45,344 : INFO : loading vocabulary recursively from word2vec3.model.vocabular
2019-10-26 22:21:45,345 : INFO : loading trainables recursively from word2vec3.model.trainable
2019-10-26 22:21:45,346 : INFO : setting ignored attribute cum_table to None
2019-10-26 22:21:45,347 : INFO : loaded word2vec3.model
In [55]: new_model[w1]
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
Out[55]: array([-0.02918519, -0.0337136 , 0.08795231, 0.02421778, 0.1780647 ,
                0.0553538, 0.12115074, 0.0975095, 0.05988046, 0.02197726,
                0.03644676, 0.10601006, -0.00516792, 0.16420509, 0.12053949,
                0.07873943, 0.05759237, -0.04323119, 0.0278474, -0.11028142,
                0.11981519, -0.0336413, 0.00403806, 0.2104455, -0.03523133,
               -0.12024543, -0.10330275, -0.11238363, -0.06754733, -0.08081181,
               -0.10363467, -0.09096432, 0.00277428, -0.00078453, 0.091377 ,
                0.20121473, -0.06839076, -0.06283233, -0.07026026, -0.08496032,
                0.05678556, 0.10152854, 0.11724606, 0.07272658, -0.12158132,
               -0.01087536, 0.11207011, 0.11630771, 0.01763994, -0.06558477],
              dtype=float32)
```

In [56]: new_model.similarity(w1, 'computer')

```
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
```

```
Out [56]: 0.536373
In [57]: new_model.wv.most_similar(positive = w1,topn=5)
2019-10-26 21:18:49,626 : INFO : precomputing L2-norms of word weight vectors
Out[57]: [('linotronic', 0.9489021301269531),
          ('decoder', 0.9417294859886169),
          ('dma', 0.9396167397499084),
          ('protocols', 0.9371689558029175),
          ('conecting', 0.9359794855117798)]
In [58]: cosine_distance (new_model, 'databases', target_list,5)
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWeets/
  import sys
Out[58]: [('linotronic', 0.94890213),
          ('decoder', 0.9417295),
          ('dma', 0.9396167),
          ('protocols', 0.93716884),
          ('conecting', 0.9359795)]
```

1.6 Changing model parameter window form 3 to 5

```
In [71]: new_model1 = Word2Vec(processed_emails, min_count=1,size= 50,workers=3, window =5, sg
        new_model1.save("word2vec4.model")
        new_model1 = Word2Vec.load("word2vec4.model")
2019-10-26 22:21:59,305 : WARNING : consider setting layer size to a multiple of 4 for greater
2019-10-26 22:21:59,306 : INFO : collecting all words and their counts
2019-10-26 22:21:59,307 : INFO : PROGRESS: at sentence #0, processed 0 words, keeping 0 word to
2019-10-26 22:21:59,540 : INFO : collected 30557 word types from a corpus of 1122850 raw words
2019-10-26 22:21:59,540 : INFO : Loading a fresh vocabulary
2019-10-26 22:21:59,594 : INFO : effective_min_count=1 retains 30557 unique words (100% of original)
2019-10-26 22:21:59,595 : INFO : effective_min_count=1 leaves 1122850 word corpus (100% of original contents)
2019-10-26 22:21:59,697 : INFO : deleting the raw counts dictionary of 30557 items
2019-10-26 22:21:59,698 : INFO : sample=0.001 downsamples 22 most-common words
2019-10-26 22:21:59,789 : INFO : estimated required memory for 30557 words and 50 dimensions:
2019-10-26 22:21:59,790 : INFO : resetting layer weights
```

```
2019-10-26 22:22:00,054 : INFO : training model with 3 workers on 30557 vocabulary and 50 feat
2019-10-26 22:22:00,834 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:22:00,842 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:22:00,843 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:22:00,843 : INFO : EPOCH - 1 : training on 1122850 raw words (1089876 effective
2019-10-26 22:22:01,677 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:22:01,691 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:22:01,692 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:22:01,693 : INFO : EPOCH - 2 : training on 1122850 raw words (1089591 effective
2019-10-26 22:22:02,467 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:22:02,473 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:22:02,474 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:22:02,475 : INFO : EPOCH - 3 : training on 1122850 raw words (1089643 effective
2019-10-26 22:22:03,329 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:22:03,339 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:22:03,341 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:22:03,342 : INFO : EPOCH - 4 : training on 1122850 raw words (1089645 effective
2019-10-26 22:22:04,077 : INFO : worker thread finished; awaiting finish of 2 more threads
2019-10-26 22:22:04,082 : INFO : worker thread finished; awaiting finish of 1 more threads
2019-10-26 22:22:04,084 : INFO : worker thread finished; awaiting finish of 0 more threads
2019-10-26 22:22:04,084 : INFO : EPOCH - 5 : training on 1122850 raw words (1089722 effective
2019-10-26 22:22:04,085 : INFO : training on a 5614250 raw words (5448477 effective words) too
2019-10-26 22:22:04,092 : INFO : saving Word2Vec object under word2vec4.model, separately None
2019-10-26 22:22:04,094 : INFO : not storing attribute vectors norm
2019-10-26 22:22:04,095 : INFO : not storing attribute cum_table
2019-10-26 22:22:04,290 : INFO : saved word2vec4.model
2019-10-26 22:22:04,291 : INFO : loading Word2Vec object from word2vec4.model
2019-10-26 22:22:04,433 : INFO : loading wv recursively from word2vec4.model.wv.* with mmap=No.
2019-10-26 22:22:04,434 : INFO : setting ignored attribute vectors norm to None
2019-10-26 22:22:04,434 : INFO : loading vocabulary recursively from word2vec4.model.vocabulary
2019-10-26 22:22:04,435 : INFO : loading trainables recursively from word2vec4.model.trainable
2019-10-26 22:22:04,436 : INFO : setting ignored attribute cum_table to None
2019-10-26 22:22:04,437 : INFO : loaded word2vec4.model
In [64]: print("Euclidean Similarity of word 'databases' to computer is: ", new_model1.similar
        print(" ")
        print("Top 5 words similar to the given word according to euclidean similarity: ",new
        print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
```

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationWesterner.py:1: DeprecationWesterner

2019-10-26 21:59:25,281 : INFO : precomputing L2-norms of word weight vectors

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationWafter removing the cwd from sys.path.

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWaimport sys

Euclidean Similarity of word 'databases' to computer is: 0.5324575

Top 5 words similar to the given word according to euclidean similarity: [('multi', 0.9284030

Top 5 words similar to the given word according to cosine similarity: [('multi', 0.9284031),

1.7 Comparison of 10 handpicked words

1.8 Model: Parameter sg = 1, window = 3

```
In [83]: print(w2)
        print("")
         print("Euclidean Similarity of word " + w2 + " to computer is: ", model.similarity(w2
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
         print(" ")
         print(w3)
         print("")
         print("Euclidean Similarity of word " + w3 + " to computer is: ", model.similarity(w3
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ", mode
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
         print(" ")
         print(w4)
         print("")
         print("Euclidean Similarity of word " + w4 + " to computer is: ", model.similarity(w4
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
         print(" ")
         print(w5)
         print("")
```

print("Euclidean Similarity of word " + w5 + " to computer is: ", model.similarity(w5

```
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w6)
print("")
print("Euclidean Similarity of word" + w6 + " to computer is: ", model.similarity(w6
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w7)
print("")
print("Euclidean Similarity of word " + w7 + " to computer is: ", model.similarity(w7
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mode
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w8)
print("")
print("Euclidean Similarity of word " + w8 + " to computer is: ", model.similarity(w8
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mode
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w9)
print("")
print("Euclidean Similarity of word " + w9 + " to computer is: ", model.similarity(w9
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mode
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w10)
print("")
print("Euclidean Similarity of word " + w10 + " to computer is: ", model.similarity(w
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mode
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
```

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:3: DeprecationWealth This is separate from the ipykernel package so we can avoid doing imports until 2019-10-26 22:43:35,166: INFO: precomputing L2-norms of word weight vectors

popular

Euclidean Similarity of word popular to computer is: 0.47004202

Top 5 words similar to the given word according to euclidean similarity: [('revision', 0.8567'

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationWater removing the cwd from sys.path.

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWindows import sys

Top 5 words similar to the given word according to cosine similarity: [('revision', 0.8567767 technology

Euclidean Similarity of word technology to computer is: 0.3986991

Top 5 words similar to the given word according to euclidean similarity: [('telecommunication

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:11: Deprecation # This is added back by InteractiveShellApp.init_path()

Top 5 words similar to the given word according to cosine similarity: [('telecommunication', question

Euclidean Similarity of word question to computer is: 0.33207655

Top 5 words similar to the given word according to euclidean similarity: [('answer', 0.818938'

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:19: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('answer', 0.8189388), information

Euclidean Similarity of word information to computer is: 0.36308715

Top 5 words similar to the given word according to euclidean similarity: [('omission', 0.7092

 $/ Users/anjaliverma/anaconda 3/lib/python 3.6/site-packages/ipykernel_launcher.py: 27: \ Deprecation 1.00 and 1.00 and$

Top 5 words similar to the given word according to cosine similarity: [('omission', 0.70929193

Euclidean Similarity of word letter to computer is: 0.23286408

Top 5 words similar to the given word according to euclidean similarity: [('morris', 0.732562

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:35: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('morris', 0.73256224) digit

Euclidean Similarity of word digit to computer is: 0.46139246

Top 5 words similar to the given word according to euclidean similarity: [('megabuck', 0.92124

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:43: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('megabuck', 0.9212495') data

Euclidean Similarity of word data to computer is: 0.37617567

Top 5 words similar to the given word according to euclidean similarity: [('acquisition', 0.7

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:51: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('acquisition', 0.76255) prohibition

```
Euclidean Similarity of word prohibition to computer is: 0.30014408

Top 5 words similar to the given word according to euclidean similarity: [('levitical', 0.889)

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:59: Deprecation
```

Euclidean Similarity of word faith to computer is: 0.1316617

Top 5 words similar to the given word according to euclidean similarity: [('reject', 0.736490

Top 5 words similar to the given word according to cosine similarity: [('levitical', 0.889396

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:67: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('reject', 0.73648995)

1.9 Model: Parameter sg = 1, window = 5

faith

```
In [84]: print(w2)
         print("")
         print("Euclidean Similarity of word " + w2 + " to computer is: ", model1.similarity(wi
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ", mode
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
         print(" ")
         print(w3)
         print("")
         print("Euclidean Similarity of word " + w3 + " to computer is: ", model1.similarity(was print("Euclidean Similarity))
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
         print(" ")
         print(w4)
         print("")
         print("Euclidean Similarity of word " + w4 + " to computer is: ", model1.similarity(w-
         print(" ")
```

print("Top 5 words similar to the given word according to euclidean similarity: ", mode

```
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w5)
print("")
print("Euclidean Similarity of word " + w5 + " to computer is: ", model1.similarity(was a similarity)
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w6)
print("")
print("Euclidean Similarity of word " + w6 + " to computer is: ", model1.similarity(w
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w7)
print("")
print("Euclidean Similarity of word " + w7 + " to computer is: ", model1.similarity(w
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mode
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w8)
print("")
print("Euclidean Similarity of word " + w8 + " to computer is: ", model1.similarity(w
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w9)
print("")
print("Euclidean Similarity of word " + w9 + " to computer is: ", model1.similarity(w
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w10)
print("")
print("Euclidean Similarity of word " + w10 + " to computer is: ", model1.similarity(
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ", mod-
```

```
print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:3: DeprecationW
  This is separate from the ipykernel package so we can avoid doing imports until
2019-10-26 22:45:27,782 : INFO : precomputing L2-norms of word weight vectors
popular
Euclidean Similarity of word popular to computer is: 0.43820477
Top 5 words similar to the given word according to euclidean similarity: [('halfway', 0.80031
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Top 5 words similar to the given word according to cosine similarity: [('halfway', 0.80031914
technology
Euclidean Similarity of word technology to computer is: 0.3719428
Top 5 words similar to the given word according to euclidean similarity: [('prodigy', 0.73267
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:11: Deprecation
  # This is added back by InteractiveShellApp.init_path()
Top 5 words similar to the given word according to cosine similarity: [('prodigy', 0.7326734)
question
Euclidean Similarity of word question to computer is: 0.21110871
Top 5 words similar to the given word according to euclidean similarity: [('answer', 0.859965
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:19: Deprecation
```

Top 5 words similar to the given word according to cosine similarity: [('answer', 0.85996604) information

Euclidean Similarity of word information to computer is: 0.4516124

Top 5 words similar to the given word according to euclidean similarity: [('herein', 0.716072

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:27: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('herein', 0.7160725), letter

Euclidean Similarity of word letter to computer is: 0.18696895

Top 5 words similar to the given word according to euclidean similarity: [('moody', 0.71299958

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:35: DeprecationN

Top 5 words similar to the given word according to cosine similarity: [('moody', 0.7129996), digit

Euclidean Similarity of word digit to computer is: 0.43333927

Top 5 words similar to the given word according to euclidean similarity: [('uncompressed', 0.5

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:43: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('uncompressed', 0.887 data

Euclidean Similarity of word data to computer is: 0.29505494

Top 5 words similar to the given word according to euclidean similarity: [('acquisition', 0.78

```
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:51: Deprecation Top 5 words similar to the given word according to cosine similarity: [('acquisition', 0.7813') prohibition

Euclidean Similarity of word prohibition to computer is: 0.22733936

Top 5 words similar to the given word according to euclidean similarity: [('rum', 0.869230508') / Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:59: Deprecation'

Top 5 words similar to the given word according to cosine similarity: [('rum', 0.8692305), ('daith') faith
```

Euclidean Similarity of word faith to computer is: 0.2571972

Top 5 words similar to the given word according to euclidean similarity: [('believer', 0.77539

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:67: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('believer', 0.7753929

1.10 Model: Parameter sg = 0, window = 3

```
In [85]: print(w2)
    print("")
    print("Euclidean Similarity of word " + w2 + " to computer is: ", new_model.similarity
    print(" ")
    print("Top 5 words similar to the given word according to euclidean similarity: ",new_print(" ")
    print("Top 5 words similar to the given word according to cosine similarity: ", cosine print(" ")
    print(w3)
    print("")
    print("Euclidean Similarity of word " + w3 + " to computer is: ", new_model.similarity
    print(" ")
    print("Top 5 words similar to the given word according to euclidean similarity: ",new_print(" ")
```

print("Top 5 words similar to the given word according to cosine similarity: ", cosine

```
print(" ")
print(w4)
print("")
print("Euclidean Similarity of word " + w4 + " to computer is: ", new_model.similarity
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w5)
print("")
print("Euclidean Similarity of word " + w5 + " to computer is: ", new_model.similarity
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w6)
print("")
print("Euclidean Similarity of word " + w6 + " to computer is: ", new_model.similarity
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w7)
print("")
print("Euclidean Similarity of word " + w7 + " to computer is: ", new_model.similarity
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w8)
print("")
print("Euclidean Similarity of word " + w8 + " to computer is: ", new_model.similarity
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
print(" ")
print(w9)
print("")
print("Euclidean Similarity of word " + w9 + " to computer is: ", new_model.similarit
print(" ")
print("Top 5 words similar to the given word according to euclidean similarity: ",new
print(" ")
print("Top 5 words similar to the given word according to cosine similarity: ", cosine
```

```
print("")
        print("Euclidean Similarity of word " + w10 + " to computer is: ", new_model.similari
        print(" ")
        print("Top 5 words similar to the given word according to euclidean similarity: ",new
        print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:3: DeprecationW
  This is separate from the ipykernel package so we can avoid doing imports until
2019-10-26 22:48:21,882 : INFO : precomputing L2-norms of word weight vectors
popular
Euclidean Similarity of word popular to computer is: 0.50274193
Top 5 words similar to the given word according to euclidean similarity: [('actively', 0.9416-
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Top 5 words similar to the given word according to cosine similarity: [('actively', 0.9416484
technology
Euclidean Similarity of word technology to computer is: 0.67519844
Top 5 words similar to the given word according to euclidean similarity: [('telecommunication
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:11: Deprecation
  # This is added back by InteractiveShellApp.init_path()
Top 5 words similar to the given word according to cosine similarity: [('telecommunication',
question
Euclidean Similarity of word question to computer is: 0.21732914
Top 5 words similar to the given word according to euclidean similarity: [('clergymen', 0.818
```

print(" ")
print(w10)

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:19: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('clergymen', 0.818915) information

Euclidean Similarity of word information to computer is: 0.62731767

Top 5 words similar to the given word according to euclidean similarity: [('correction', 0.82'

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:27: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('correction', 0.82753

Euclidean Similarity of word letter to computer is: 0.20623244

Top 5 words similar to the given word according to euclidean similarity: [('suggestion', 0.83

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:35: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('suggestion', 0.83224: digit

Euclidean Similarity of word digit to computer is: 0.38559967

Top 5 words similar to the given word according to euclidean similarity: [('lite', 0.98671293

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:43: Deprecation

Top 5 words similar to the given word according to cosine similarity: [('lite', 0.986713), ('data

```
Euclidean Similarity of word data to computer is: 0.425191
Top 5 words similar to the given word according to euclidean similarity: [('transfer', 0.8559
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:51: Deprecation
Top 5 words similar to the given word according to cosine similarity: [('transfer', 0.8559536
prohibition
Euclidean Similarity of word prohibition to computer is: 0.4032716
Top 5 words similar to the given word according to euclidean similarity: [('outright', 0.9688
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:59: Deprecation
Top 5 words similar to the given word according to cosine similarity: [('outright', 0.9688137
faith
Euclidean Similarity of word faith to computer is: 0.15602455
Top 5 words similar to the given word according to euclidean similarity: [('salvation', 0.864
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:67: Deprecation
Top 5 words similar to the given word according to cosine similarity: [('salvation', 0.864400
1.11 Model: Parameter sg = 0, window = 5
In [74]: print("Euclidean Similarity of word " + w2 + " to computer is: ", new_model1.similari
```

```
print("Euclidean Similarity of word " + w2 + " to computer is: ", new_modell.similarity
    print(" ")
    print("Top 5 words similar to the given word according to euclidean similarity: ",new_print(" ")
    print("Top 5 words similar to the given word according to cosine similarity: ", cosine.")
```

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationWa

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationWe

```
print("Top 5 words similar to the given word according to euclidean similarity: ",new
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word technology to computer is: 0.6641607
Top 5 words similar to the given word according to euclidean similarity: [('specialize', 0.87
Top 5 words similar to the given word according to cosine similarity: [('specialize', 0.87839
In [76]: print("Euclidean Similarity of word " + w4 + " to computer is: ", new_model1.similari
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ",new
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
```

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWeets/

Top 5 words similar to the given word according to euclidean similarity: [('worthwhile', 0.93-

Top 5 words similar to the given word according to cosine similarity: [('worthwhile', 0.93453

In [75]: print("Euclidean Similarity of word " + w3 + " to computer is: ", new model1.similari

after removing the cwd from sys.path.

Euclidean Similarity of word popular to computer is: 0.48639983

import sys

print(" ")

Euclidean Similarity of word question to computer is: 0.061645284

```
In [77]: print("Euclidean Similarity of word" + w5 + " to computer is: ", new_model1.similari
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ",new
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationWeets/
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word information to computer is: 0.5705979
Top 5 words similar to the given word according to euclidean similarity: [('corrects', 0.8025)
Top 5 words similar to the given word according to cosine similarity: [('corrects', 0.8025624
In [78]: print("Euclidean Similarity of word " + w6 + " to computer is: ", new_model1.similari
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ",new
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word letter to computer is: 0.10323852
Top 5 words similar to the given word according to euclidean similarity: [('nico', 0.78701168
Top 5 words similar to the given word according to cosine similarity: [('nico', 0.7870118), (
In [79]: print("Euclidean Similarity of word " + w7 + " to computer is: ", new_model1.similari
```

Top 5 words similar to the given word according to euclidean similarity: [('adnausium', 0.730

Top 5 words similar to the given word according to cosine similarity: [('adnausium', 0.730890

print(" ")

```
print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationWeets/
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word digit to computer is: 0.2912973
Top 5 words similar to the given word according to euclidean similarity: [('rx', 0.9688238501
Top 5 words similar to the given word according to cosine similarity: [('rx', 0.9688239), ('m
In [80]: print("Euclidean Similarity of word" + w8 + " to computer is: ", new model1.similari
        print(" ")
        print("Top 5 words similar to the given word according to euclidean similarity: ",new
        print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word data to computer is: 0.40867683
Top 5 words similar to the given word according to euclidean similarity: [('transmission', 0.3
Top 5 words similar to the given word according to cosine similarity: [('transmission', 0.870
In [82]: print("Euclidean Similarity of word" + w9 + " to computer is: ", new_model1.similari
        print(" ")
        print("Top 5 words similar to the given word according to euclidean similarity: ",new
        print(" ")
        print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
```

print("Top 5 words similar to the given word according to euclidean similarity: ",new

/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationW

"""Entry point for launching an IPython kernel.

```
after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationWeets/
  import sys
Euclidean Similarity of word prohibition to computer is: 0.34777611
Top 5 words similar to the given word according to euclidean similarity: [('observation', 0.9
Top 5 words similar to the given word according to cosine similarity: [('observation', 0.9675
In [81]: print("Euclidean Similarity of word " + w10 + " to computer is: ", new_model1.similar
         print(" ")
         print("Top 5 words similar to the given word according to euclidean similarity: ",new
         print(" ")
         print("Top 5 words similar to the given word according to cosine similarity: ", cosine
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: DeprecationW
  """Entry point for launching an IPython kernel.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:4: DeprecationWeets/
  after removing the cwd from sys.path.
/Users/anjaliverma/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:7: DeprecationW
  import sys
Euclidean Similarity of word faith to computer is: -0.0008394681
Top 5 words similar to the given word according to euclidean similarity: [('truth', 0.8152226
Top 5 words similar to the given word according to cosine similarity: [('truth', 0.8152227),
In []:
```