### **WEEK 12 ASSIGNMENT**

#### PYTHON:

```
import pandas as pd
import numpy as np
import string
from nltk import pos_tag
from time import time
from nltk.tokenize import word_tokenize
from nltk.stem.snowball import SnowballStemmer
from nltk.stem import WordNetLemmatizer from nltk.corpus import wordnet as wn from nltk.corpus import stopwords
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.decomposition import LatentDirichletAllocation
from sklearn.decomposition import TruncatedSVD
from sklearn.decomposition import NMF
import matplotlib.pyplot as plt
from wordcloud import WordCloud, STOPWORDS from PIL import Image
import random
pd.set_option('max_colwidth', 32000)
# Read N=13,575 Hotel Reviews
df = pd.read_excel("hotels.xlsx")
sentiment_file = pd.read_excel("afinn_sentiment_words.xlsx")
def my_analyzer(s):
    s = s.lower()
    # Replace special characters with spaces
    s = s.replace('-', ' ')
s = s.replace('_', ' ')
s = s.replace(',', '.')
    # Replace not contraction with not
    s = s.replace("'nt", " not")
s = s.replace("n't", " not")
    tokens = word_tokenize(s)
    and (word != 'object') and (word != "'s")]
    for i in range(len(tokens)):
    if tokens[i] in syns:
             tokens[i] = syns[tokens[i]]
    punctuation = list(string.punctuation) + ['...', '....']
```

```
and (not word.replace("'", '', 2).isdigit())]
    # Lemmatization & Stemming - Stemming with WordNet POS
    # Since lemmatization requires POS need to set POS
    tag_wrds = pos_tag(filtered_terms, lang='eng')
# Stemming with for terms without WordNet POS
stemmer = SnowballStemmer("english")
wn_tags = {'N': wn.NOUN, 'J': wn.ADJ, 'V': wn.VERB, 'R': wn.ADV}
    wnl = WordNetLemmatizer()
    stemmed_tokens = []
    for tagged_token in tag_wtag:
        term = tagged_token[0]
        pos = tagged_token[1]
        pos = pos[0]
            pos = wn_tags[pos]
            stemmed_tokens.append(wnl.lemmatize(term, pos=pos))
            stemmed tokens.append(stemmer.stem(term))
    return stemmed tokens
def display_topics(lda, terms, n_terms=15):
    for topic_idx, topic in enumerate(lda):
        if topic idx > 8:
            break
        message = "Topic #%d: " % (topic idx + 1)
        print(message)
        abs_topic = abs(topic)
topic_terms_sorted = \
    [[terms[i], topic[i]] \
              for i in abs_topic.argsort()[:-n_terms - 1:-1]]
        n = int(n_terms / k)
        m = n_{terms} - k * n
            message = ''
for i in range(k):
                 if topic_terms_sorted[i + l][1] > 0:
                     word = "+" + topic_terms_sorted[i + l][0]
                 word = "-" + topic_terms_sorted[i + l][0]
message += '{:<15s}'.format(word)</pre>
            print(message)
            l = k * n
            message = ''
            for i in range(m):
                 if topic_terms_sorted[i + l][1] > 0:
                     word = "+" + topic_terms_sorted[i + l][0]
                     word = "-" + topic_terms_sorted[i + l][0]
                message += '{:<15s}'.format(word)</pre>
            print(message)
```

```
attribute_map = {
      'doc': [3, (0, 1e+12), [0, 0]],
'hotel': [2, ('Bally's', 'Bellagio', 'Circus Circus', 'Encore', 'Excalibur'),
[0, 0]],
     0]],
'Review': [3, (''), [0, 0]],
'topic': [2, (0, 1, 2, 3, 4, 5, 6), [0, 0]],
'T1': [0, (-1e+8, 1e+8), [0, 0]],
'T2': [0, (-1e+8, 1e+8), [0, 0]],
'T3': [0, (-1e+8, 1e+8), [0, 0]],
'T4': [0, (-1e+8, 1e+8), [0, 0]],
'T5': [0, (-1e+8, 1e+8), [0, 0]],
'T6': [0, (-1e+8, 1e+8), [0, 0]],
'T7': [0, (-1e+8, 1e+8), [0, 0]],
etup program constants
n_comments = len(df['Review']) # Number of hotel reviews
m_features = None # Number of SVD Vectors
stop_words = 'english' # Stop Word Dictionary
comments = df['Review'] # place all text reviews in reviews
num_topics = 7  # number of topic clusters to extract
num_iterations = 10  # maximum number of itertions
max_df = 0.7 # learning offset for LDAmax proportion of docs/reviews allowed for a
tf = cv.fit_transform(comments)
terms = cv.get_feature_names()
term_sums = tf.sum(axis=0)
term counts = []
for i in range(len(terms)):
     term_counts.append([terms[i], term_sums[0, i]])
def sortSecond(e):
     return e[1]
term_counts.sort(key=sortSecond, reverse=True)
print("\nTerms with Highest Frequency:")
for i in range(10):
     print('{:<15s}{:>5d}'.format(term_counts[i][0], term_counts[i][1]))
# Modify tf, term frequencies, to TF/IDF matrix from the data print("Conducting Term/Frequency Matrix using TF-IDF")

tfidf_vect = TfidfTransformer(norm=None, use_idf=True) # set norm=None
tf = tfidf_vect.fit_transform(tf)
term_idf_sums = tf.sum(axis=0)
term_idf_scores = []
term_idf_scores.sort(key=sortSecond, reverse=True)
print("\nTerms with Highest TF-IDF Scores:")
for i in range(10):
     print('{:<15s}{:>8.2f}'.format(term_idf_scores[i][0], \
term_idf_scores[i][1]))
num_reviews = len(df['Review']) # Number of wine reviews
m_features = 100 # Number of SVD Vectors
stop_words = 'english' # Stop Word Dictionary
ngram = (1, 2) \# n-gram POS modeling
reviews = df['Review'] # place all text reviews in reviews
num_topics = 9 # number of topic clusters to extract
```

```
num_iterations = 10  # maximum number of itertions for LDA
learning offset = 10. # learning offset for LDA
learning method = 'online' # learning method for LDA
tfidf = True
lda = LatentDirichletAllocation(n_components=num_topics,
 num_iterations=num_iterations, \
                                              learning_method=learning_method, \
learning_offset=learning_offset, \
random_state=12345)
U = lda.fit_transform(tf)
print('{:.<22s}{:>6d}'.format("Number of Reviews", len(reviews)))
print('{:.<22s}{:>6d}'.format("Number of Terms", len(terms)))
print("\nTopics Identified using LDA with TF_IDF")
display_topics(lda.components_, terms, n_terms=15)
# Store topic group for each doc in topics[]
topics = [0] * n_comments
topic_counts = [0] * (num_topics + 1)
for i in range(n_comments):
      max = abs(U[i][0])
      topics[i] = 0
      for j in range(num_topics):
    x = abs(U[i][j])
                 topics[i] = i
      topic_counts[topics[i]] += 1
print('{:<6s}{:>8s}'.format("TOPIC", "COMMENTS", "PERCENT"))
for i in range(num_topics):
                                                           topic_counts[i] / n_comments))
comment_scores = []
for i in range(n_comments):
      u = [0] * (num_topics + 1)
u[0] = topics[i]
      for j in range(num_topics):
            u[i + 1] = U[i][i]
      comment_scores.append(u)
 for i in range(num_topics):
      cols.append(s)
df_topics = pd.DataFrame.from_records(comment_scores, columns=cols)
df = df.join(df_topics)
sentiment_dic = {}
for i in range(len(sentiment_file)):
    sentiment_dic[sentiment_file.iloc[i][0]] = sentiment_file.iloc[i][1]
def my_preprocessor(s):
      s = s.lower()
     s = s.replace('-', '')
s = s.replace('_', '')
s = s.replace(',', '.')
     # Replace not contraction with not
s = s.replace("'nt", " not")
s = s.replace("n't", " not")
```

```
preprocessor=my_preprocessor, ngram_range=(1, 2))
tf = cv.fit_transform(df['Review'])
terms = cv.get_feature_names()
num_reviews = tf.shape[0]
n terms = tf.shape[1]
print('{:.<22s}{:>6d}'.format("Number of Reviews", num_reviews))
print('{:.<22s}{:>6d}'.format("Number of Terms", n_terms))
min sentiment = +5
max_sentiment = -5
avg_sentiment, min, max = 0, 0, 0
min_list, max_list = [], []
sentiment_score = [0] * num_reviews
for i in range(num_reviews):
    term_list = tf[i].nonzero()[1]
    if len(term_list) > 0:
        for t in np.nditer(term_list):
            score = sentiment_dic.get(terms[t])
            if score != None:
                sentiment_score[i] += score * tf[i, t]
                n_sentiment_file += tf[i, t]
    if n_sentiment_file > 0:
        sentiment_score[i] = sentiment_score[i] / n_sentiment_file
    if sentiment_score[i] == max_sentiment and n_sentiment_file > 3:
       max list.append(i)
    if sentiment_score[i] > max_sentiment and n_sentiment_file > 3:
        max_sentiment = sentiment_score[i]
        max = i
       max list = [i]
    if sentiment_score[i] == min_sentiment and n_sentiment_file > 3:
       min_list.append(i)
    if sentiment_score[i] < min_sentiment and n_sentiment_file > 3:
        min_sentiment = sentiment_score[i]
        min_list = [i]
    avg_sentiment += sentiment_score[i]
avg_sentiment = avg_sentiment / num_reviews
print("\nCorpus Average Sentiment: ", avg_sentiment)
print("\nMost Negative Reviews with 4 or more Sentiment Words:")
for i in range(len(min_list)):
    ', min_sentiment))
print("\nMost Positive Reviews with 4 or more Sentiment Words:")
for i in range(len(max_list)):
                                            " Sentiment is ", max_sentiment))
stopw = set(STOPWORDS)
stopw.add("stay")
stopw.add("hotel")
stopw.add("room")
stopw.add("consider")
def shades_of_grey(word, font_size, position, orientation, random_state=None,
    return "hsl(0, 0%, %d%)" % random.randint(60, 1000)
```

```
cloud.generate(df['Review'].iloc[min_list[0]])
print("\nMost Negative Review Sentiment=", min_sentiment)
plt.imshow(cloud.recolor(color_func=shades_of_grey, random_state=3), \
plt.axis("off")
plt.figure()
i = min_list[0]
term_list = tf[i].nonzero()[1]
if len(term_list) > 0:
    for t in np.nditer(term list):
        score = sentiment_dic.get(terms[t])
        if score != None:
    s += terms[t] + " "
cloud.generate(s)
plt.imshow(cloud.recolor(color_func=shades_of_grey, random_state=3), \
plt.axis("off")
plt.figure()
plt.show()
cloud.generate(df['Review'].iloc[max_list[0]])
plt.axis("off")
plt.figure()
i = max list[0]
term_list = tf[i].nonzero()[1]
if len(term_list) > 0:
    for t in np.nditer(term list):
        score = sentiment_dic.get(terms[t])
            s += terms[t] + " "
cloud.generate(s)
plt.imshow(cloud.recolor(color_func=shades_of_grey, random_state=3),
plt.axis("off")
plt.figure()
plt.show()
corpus_sentiment = {}
n_sentiment_file = 0
for i in range(num_reviews):
    # Iterate over the terms with nonzero scores
    term_list = tf[i].nonzero()[1]
    if len(term_list) > 0:
        for t in np.nditer(term_list):
            score = sentiment_dic.get(terms[t])
            if score != None:
    n_sentiment_file += tf[i, t]
                current_count = corpus_sentiment.get(terms[t])
                if current_count == None:
                    corpus_sentiment[terms[t]] = tf[i, t]
                    corpus_sentiment[terms[t]] += tf[i, t]
print("The Corpus contains a total of ", len(corpus_sentiment), " unique sentiment
print("The total number of sentiment words in the Corpus is", n_sentiment_file)
cloud.generate_from_frequencies(corpus_sentiment)
plt.imshow(cloud.recolor(color_func=shades_of_grey, random_state=3),
plt.axis("off")
```

## plt.figure() plt.show()

```
OUTPUT:
Terms with Highest Frequency:
                1603
great
                1494
go
good
                1432
would
                1406
strip
                1243
                1202
time
check
                1191
night
                1179
one
                1162
quot
                1162
Conducting Term/Frequency Matrix using TF-IDF
('The Term/Frequency matrix has', 1671, 'rows, and', 3152, 'columns.'
('The Term list has', 3152, 'terms.')
Terms with Highest TF-IDF Scores:
quot
                2962.70
great
                2721.80
                2602.81
go
                2564.84
good
would
                2458.36
check
                2281.95
strip
                2247.83
                2234.49
pool
time
                2223.09
one
                2212.57
Number of Reviews.... 1671
Number of Terms..... 3152
Topics Identified using LDA with TF IDF
Topic #1:
+circus
               +kid
                               +quot
                                              +qo
                                                              +bad
+get
               +one
                               +like
                                              +family
                                                              +place
               +excalibur
+take
                               +people
                                              +nice
                                                              +pay
Topic #2:
+quot
               +luggage
                               +guard
                                              +husband
                                                             +daughter
+help
               +security
                               +bag
                                              +order
                                                             +make
+cart
               +wheelchair
                               +wait
                                              +shuttle
                                                             +dinner
Topic #3:
+smoke
               +non
                               +call
                                              +check
                                                             +would
+smoking
                                              +circus
               +day
                               +tell
                                                             +charge
+book
                               +per
                                                             +quot
               +even
                                              +go
Topic #4:
+bellagio
               +great
                               +pool
                                              +view
                                                              +best
                                              +fountain
+love
               +strip
                               +qood
                                                             +night
+excellent
                               +location
                                              +would
                                                             +beautiful
               +vega
Topic #5:
+pool
               +hot
                                                              +day
                               +york
                                              +new
```

Name: Krit Gupta UIN: 927001565				
+water +couple	+cold +need	+great +look	+inside +place	+fresh +food
_	· need	Look	Pidoc	11000
Topic #6: +excalibur	+value	± amp	+good	+x003
+time	+tower	+amp +strip	+good +money	+vega +bally
+clean	+great	+circus	+spend	+location
relean	rgreat	refreus	rapend	1100401011
Topic #7:				
+encore	+wynn	+strip	+great	+mandalay
+area	+walk	+restaurant	+bay	+casino
+bellagio	+pool	+cafe	+nice	+shop
Topic #8:				
+encore	+wynn	+suite	+w/	+service
+check	+guest	+bed	+floor	+area
+quot	+spa	+light	+give	+beautiful
Topic #9:				
+check	+would	+quot	+ask	+qo
+night	+kid	+back	+bed	+sheet
+get	+take	+desk	+could	+upgrade
TOPIC COMMENTS PERCENT				
1 290	0.0%			
2 34	0.0%			
3 165	0.0%			
4 386	0.0%			
5 110	0.0%			
6 217	0.0%			
7 150	0.0%			
8 152	0.0%			
9 167	0.0%			
Number of Reviews 1671				
Number of Terms 17527				
('\nCorpus Average Sentiment: ', 1.3219487240725591)				
Most Negative Reviews with 4 or more Sentiment Words: Review 57 Sentiment is -2.75				
Most Positive Reviews with 4 or more Sentiment Words:				

### WORDCLOUDS:

Review 628 Sentiment is 3.75 Review 1236 Sentiment is 3.75

('\nMost Negative Review Sentiment=', -2.75)



# worst baddirty

('\nMost Positive Review Sentiment=', 3.75)



# good fantastic

('The Corpus contains a total of ', 485, ' unique sentiment words') ('The total number of sentiment words in the Corpus is', 22524)

