

Natural Language Processing
(IST-664 2022-1005)

Homework Assignment: #2
Student Name: Ryan Tervo

ASSIGNMENT OBJECTIVE:

- Conduct sentiment analysis from the Homework #1 texts.
- The first homework assignment texts were 'melville-moby_dick.txt' and 'carroll-alice.txt'.

PART 1: DESCRIPTIVE STATISTICS

- To gain understanding of the sentiment you will extract those sentences that contain adjective or adverb phrases.
- Once you extract these sentences, you will do a descriptive statistics about them, e.g.,
 - Average length of a sentence
 - Top 50 adjective/adverb phrases (by frequency)
 - Top 50 adjective/adverb words
 - Top 50 nouns or verbs
- You are encouraged to combine these findings with the other fields of the text (ngrams, frequency of POS tags by group, etc).
 - You can use tables or figures to show these summaries.

PART 2: INTERPRETATION OF THE RESULTS:

- Provide interpretation of the Part 1 results.
- What do you learn from these results?

PART 3: VISION FOR THE FUTURE

- The last thing to do is to envision how you would conduct a sentiment analysis on the review texts.

SUBMISSION REQUIREMENTS:

- Use Jupyter Notebook to develop a report to explain your processing and analysis process.
- Integrate all of your Python codes, outputs, and detailed writeup explaining your workflow and interpretation in the notebook and export the report as a HTML file for submission.
- As always, please explain your process – both for cleaning and for analysis.
- Each student should submit a HTML report exported from Jupyter Notebook that includes your Python codes, outputs, and detailed writeup explaining your workflow and interpretation of the results.
- How to Submit this Homework:
 - Go to the 2U system and the Assignment for Homework #2 and attach your report.

Your submission should include:

1. A HTML report with description of the cleaning and analysis processes and results of the analysis and interpretation of the results, including tables/figures.
2. Include your thoughts on how to conduct sentiment analysis in the above report.

```
In [1]: # PART 0: INITIAL SETUP
#####
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#   IMPORT LIBRARIES:
import nltk
from nltk import FreqDist
import re
import pandas as pd

#   DEFINE FUNCTIONS:
#####
def TagFunction(grammar_re, taggedtext):
    grammar_label = grammar_re.split(':')[0]

    #   PARSE EACH SENTENCE:
    chunk_parser = nltk.RegexpParser(grammar_re)

    grammar_tags = []
    for sent in taggedtext:
        if len(sent) > 0:
            tree = chunk_parser.parse(sent)
            for subtree in tree.subtrees():
                if subtree.label() == grammar_label:
                    grammar_tags.append(subtree)

    #   STORE PHRASES:
    grammar_phrases = []
    for sent in grammar_tags:
        temp = ''
        for w, t in sent:
            temp += w + ' '
        grammar_phrases.append(temp)

    #   FREQUENCY DISTRIBUTIONS
    freq_grammar = nltk.FreqDist(grammar_phrases)
    return grammar_tags, grammar_phrases, freq_grammar

#   DEFINE VARIABLES:
#####
#   Select File Names out of nltk corpus
fileName1 = nltk.corpus.gutenberg.fileids()[7]
fileName2 = nltk.corpus.gutenberg.fileids()[12]

#   Gather text for those two files.
fileText1 = nltk.corpus.gutenberg.raw(fileName1)
fileText2 = nltk.corpus.gutenberg.raw(fileName2)

#   Determine the number of tokens in each text.
fileTextLength1 = len(nltk.word_tokenize(fileText1))
fileTextLength2 = len(nltk.word_tokenize(fileText2))

#   Print File Names to ensure the write text was selected.
print('For for the first text I selected was "' + fileName1 + '" and it has ' +
      str(fileTextLength1) + ' tokens prior to any processing.')
print('For for the second text I selected was "' + fileName2 + '" and it has ' +
      str(fileTextLength2) + ' tokens prior to any processing.')

```

For for the first text I selected was "carroll-alice.txt" and it has 33494 tokens prior to any processing.

For for the second text I selected was "melville-moby_dick.txt" and it has 255028 tokens prior to any processing.

In [2]: #####;

PART 1: DESCRIPTIVE STATISTICS

To gain understanding of the sentiment you will extract those sentences that contain adjective or adverb phrases. Once you extract these sentences, you will do a descriptive statistics about them, e.g.,

- Average length of a sentence
- Top 50 adjective/adverb phrases (by frequency)
- Top 50 adjective/adverb words
- Top 50 nouns or verbs

You are encouraged to combine these findings with the other fields of the text (ngrams, frequency of POS tags by group, etc).

- You can use tables or figures to show these summaries.

```
In [3]: # PART 1.1: TOKENIZE:
# SPLIT TEXT INTO SENTENCES:
textsplt1 = nltk.sent_tokenize(fileText1)
textsplt2 = nltk.sent_tokenize(fileText2)

# APPLY WORD TOKENIZER TO EACH SENTENCE:
tokentext1 = [nltk.word_tokenize(sent) for sent in textsplt1]
tokentext2 = [nltk.word_tokenize(sent) for sent in textsplt2]

# TAG TEXT:
taggedtext1 = [nltk.pos_tag(tokens) for tokens in tokentext1]
taggedtext2 = [nltk.pos_tag(tokens) for tokens in tokentext2]

# DEFINE RE TAGS
grammar_adjph = "ADJPH: {<RB.?>+<JJ.?>}"
grammar_advph = "ADVPH: {<RB>+<RB>}"
#grammar_other = "ahhaha"

# RUN FREQUENCY DISTRIBUTION FUNCTION:
grammar_adjph_tags1, grammar_adjph_phrases1, grammar_adjph_freqDist1 =
TagFunction(grammar_adjph, taggedtext1)
grammar_advph_tags1, grammar_advph_phrases1, grammar_advph_freqDist1 =
TagFunction(grammar_advph, taggedtext1)

# RUN FREQUENCY DISTRIBUTION FUNCTION:
grammar_adjph_tags2, grammar_adjph_phrases2, grammar_adjph_freqDist2 =
TagFunction(grammar_adjph, taggedtext2)
grammar_advph_tags2, grammar_advph_phrases2, grammar_advph_freqDist2 =
TagFunction(grammar_advph, taggedtext2)

In [4]: #-----

In [5]: # PART 1.2a: IDENTIFY
#####
print('For for the first text I selected was "' + fileName1)
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```

# TOP 50 ADJECTIVE PHRASES
#####
print('\nText1 - Top adjective phrases by frequency: \n' + '-'*75)
for word, freq in grammar_adjph_freqDist1.most_common(50):
    if len(word) <= 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1

    print(word, '\t'*tabMult, freq)
print('Length of adjective phrase sentences: ', len(grammar_adjph_tags1))

# TOP 50 ADVERB PHRASES
#####
print('\nText1 - Top adverb phrases by frequency: \n' + '-'*75)
for word, freq in grammar_advph_freqDist1.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)
print('Length of adverb phrase sentences: ', len(grammar_advph_tags1))

# TOP 50 ADJECTIVE TOKENS:
#####
adjective_tokens1 = []
for sentence in taggedtext1:
    for word, pos in sentence:
        if pos in ['JJ', 'JJR', 'JJS']: # adjective, comparative, superlative
            if len(word) > 1:
                adjective_tokens1.append(word)
freq_adjective1 = nltk.FreqDist(adjective_tokens1)

print('\nText1 - Top 50 Adjective Tokens\n' + '-'*75 )
for word, freq in freq_adjective1.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 ADVERB TOKENS:
#####
adverb_tokens1 = []
for sentence in taggedtext1:
    for word, pos in sentence:
        if pos in ['RB', 'RBR', 'RBS']: # adverb, comparative, superlative
            if len(word)>1:

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        adverb_tokens1.append(word)
freq_adverb1 = nltk.FreqDist(adverb_tokens1)

print('\nText1 - Top 50 Adverb Tokens\n' + '-'*75 )
for word, freq in freq_adverb1.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 NOUN TOKENS:
#####
noun_tokens1 = []
for sentence in taggedtext1:
    for word, pos in sentence:
        if pos in ['NN', 'NNS', 'NNP', 'NNPS']: # noun, noun plural, proper noun,
proper noun plural
            if len(word) > 1:
                noun_tokens1.append(word)
freq_noun1 = nltk.FreqDist(noun_tokens1)

print('\nText1 - Top 50 Noun Tokens\n' + '-'*75 )
for word, freq in freq_noun1.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 VERB TOKENS:
#####
verb_tokens1 = []
for sentence in taggedtext1:
    for word, pos in sentence:
        if pos in ['VB', 'VBD', 'VBG', 'VBN', 'VBP', 'VBZ']: # verb, verb past tense,
verb gerund, verb past principle, vrb non 3rd person singular, verb 3rd person
singular present ural
            if len(word) > 1:
                verb_tokens1.append(word)
freq_verb1 = nltk.FreqDist(verb_tokens1)

print('\nText1 - Top 50 Verb Tokens\n' + '-'*75 )
for word, freq in freq_verb1.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

```

```
#####
# Now we have two lists of POS tags combinations we can compare
# We need to get the sentences back from the tagging exercise and run some stats

# Create a list of original sentences from the ADJECTIVE phrase subset:
adjph_whole_sentences1 = []

# loop over the sentences in the adjective phrase sentences we created:
for sent in grammar_adjph_tags1:
    temp=''
    for (word, tag) in sent:
        temp += word+' '
    adjph_whole_sentences1.append(temp)

print('\n', '=' * 50, '\n', len(adjph_whole_sentences1))
```

For the first text I selected was "carroll-alice.txt"

Text1 - Top adjective phrases by frequency:

so much	8
very curious	6
very glad	5
very much	4
very little	4
very likely	3
so many	3
too much	3
very tired	2
quite natural	2
very deep	2
n't much	2
very few	2
very good	2
as much	2
so grave	2
always ready	2
very uncomfortable	2
almost wish	2
very difficult	2
quite silent	2
certainly too much	2
not much	2
so large	2
n't very civil	2
very interesting	2
once more	2
very sleepy	1
so VERY remarkable	1
so VERY much	1
too dark	1
VERY good	1
rather glad	1
no longer	1
too large	1
too small	1

not much larger	1
really impossible	1
almost certain	1
very nice	1
now only ten	1
too slippery	1
very fond	1
very small	1
quite surprised	1
quite dull	1
now more	1
so desperate	1
very hot	1
ever so many	1
Length of adjective phrase sentences:	222

Text1 - Top adverb phrases by frequency:

as well	15
very soon	6
very politely	5
down here	4
very much	4
just as well	4
so VERY	3
very well	3
so far	3
Just then	3
back again	3
well enough	3
down again	3
as soon	3
very carefully	3
As soon	3
so often	3
not quite	3
very nearly	3
n't quite	3
as long	3
just now	3
very slowly	2
very earnestly	2
not even	2
too far	2
as hard	2
rather not	2
very gravely	2
very humbly	2
very angrily	2
so easily	2
certainly too	2
rather timidly	2
'All right	2
n't very	2
'Exactly so	2
asleep again	2
never even	2

n't even	2
never before	1
never once	1
suddenly down	1
not much	1
not here before	1
too long	1
VERY deeply	1
now only	1
not possibly	1
quite plainly	1
Length of adverb phrase sentences:	235

Text1 - Top 50 Adjective Tokens

little	124
other	40
great	39
much	34
large	33
last	32
more	31
first	31
such	26
poor	25
thought	24
good	24
long	23
same	23
curious	19
sure	19
next	18
old	17
right	16
low	14
high	14
whole	13
mad	13
many	12
glad	11
own	10
small	10
few	9
best	9
different	9
least	9
afraid	8
white	8
ready	8
dear	8
beautiful	8
golden	7
larger	7
enough	7
deep	6
nice	6
dry	6

bright	6
melancholy	6
offended	6
full	6
sharp	6
hot	5
likely	5
nervous	5

Text1 - Top 50 Adverb Tokens

n't	203
not	128
very	126
so	91
again	83
then	72
quite	48
now	47
as	45
just	44
never	41
only	41
here	39
down	36
once	31
well	31
back	31
too	25
rather	25
soon	24
up	24
away	23
yet	21
ever	20
even	17
much	17
more	16
indeed	15
perhaps	14
anxiously	14
hastily	14
first	13
However	13
certainly	13
far	13
suddenly	12
there	12
still	12
about	12
always	12
else	11
hardly	11
enough	11
really	10
nearly	10
So	9

Then	9
angrily	9
together	9
timidly	9

Text1 - Top 50 Noun Tokens

Alice	390
Queen	72
time	65
King	60
Turtle	58
Mock	56
Hatter	55
Gryphon	54
way	53
head	50
thing	49
voice	47
Rabbit	44
Duchess	42
tone	40
Dormouse	39
March	34
moment	31
'It	31
Hare	31
nothing	30
things	30
door	30
Mouse	29
eyes	28
Caterpillar	27
day	25
course	25
Cat	25
'You	24
round	23
White	22
words	21
minute	21
sort	20
feet	19
anything	19
hand	19
dear	18
house	18
use	17
question	17
side	17
table	17
something	17
jury	17
court	17
garden	16
end	15
idea	15

Text1 - Top 50 Verb Tokens

said	456
was	361
had	183
be	145
do	113
's	105
is	104
know	87
were	86
went	83
have	80
did	74
see	66
began	58
'm	54
say	51
think	49
thought	47
go	46
looked	45
got	45
get	44
are	42
came	40
herself	38
've	38
been	38
're	36
made	30
found	30
looking	30
put	29
replied	28
seemed	27
going	27
make	27
heard	26
come	25
tell	25
took	23
felt	23
added	23
getting	22
like	22
find	21
does	20
tried	19
being	19
take	18
spoke	17

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464

In [6]: #-----

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In [7]: # PART 1.2b: IDENTIFY
#####
print('For for the second text I selected was "' + fileName2)

# TOP 50 ADJECTIVE PHRASES
#####
print('\nText2 - Top adjective phrases by frequency: \n' + '-'*75)
for word, freq in grammar_adjph_freqDist2.most_common(50):
    if len(word) <= 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1

    print(word, '\t'*tabMult, freq)
print('Length of adjective phrase sentences: ', len(grammar_adjph_tags2))

# TOP 50 ADVERB PHRASES
#####
print('\nText2 - Top adverb phrases by frequency: \n' + '-'*75)
for word, freq in grammar_advph_freqDist2.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)
print('Length of adverb phrase sentences: ', len(grammar_advph_tags2))

# TOP 50 ADJECTIVE TOKENS:
#####
adjective_tokens2 = []
for sentence in taggedtext2:
    for word, pos in sentence:
        if pos in ['JJ', 'JJR', 'JJS']: # adjective, comparative, superlative
            if len(word) > 1:
                adjective_tokens2.append(word)
freq_adjective2 = nltk.FreqDist(adjective_tokens2)

print('\nText2 - Top 50 Adjective Tokens\n' + '-'*75 )
for word, freq in freq_adjective2.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 ADVERB TOKENS:
#####

```

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adverb_tokens2 = []
for sentence in taggedtext2:
    for word, pos in sentence:
        if pos in ['RB', 'RBR', 'RBS']: # adverb, comparative, superlative
            if len(word) > 1:
                adverb_tokens2.append(word)
freq_adverb2 = nltk.FreqDist(adverb_tokens2)

print('\nText2 - Top 50 Adverb Tokens\n' + '-'*75 )
for word, freq in freq_adverb2.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 NOUN TOKENS:
#####
noun_tokens2 = []
for sentence in taggedtext2:
    for word, pos in sentence:
        if pos in ['NN', 'NNS', 'NNP', 'NNPS']: # noun, noun plural, proper noun,
proper noun plural
            if len(word) > 1:
                noun_tokens2.append(word)
freq_noun2 = nltk.FreqDist(noun_tokens2)

print('\nText2 - Top 50 Noun Tokens\n' + '-'*75 )
for word, freq in freq_noun2.most_common(50):
    if len(word) < 7:
        tabMult = 3
    elif len(word) <= 14:
        tabMult = 2
    else:
        tabMult = 1
    print(word, '\t'*tabMult, freq)

# TOP 50 VERB TOKENS:
#####
verb_tokens2 = []
for sentence in taggedtext2:
    for word, pos in sentence:
        if pos in ['VB', 'VBD', 'VBG', 'VBN', 'VBP', 'VBZ']: # verb, verb past tense,
verb gerund, verb past principle, vrb non 3rd person singular, verb 3rd person
singular present ural
            if len(word) > 1:
                verb_tokens2.append(word)
freq_verb2 = nltk.FreqDist(verb_tokens2)

print('\nText2 - Top 50 Verb Tokens\n' + '-'*75 )
for word, freq in freq_verb2.most_common(50):
    if len(word) < 7:
        tabMult = 3

```

```

elif len(word) <= 14:
    tabMult = 2
else:
    tabMult = 1
print(word, '\t'*tabMult, freq)

```

```

#####
# Now we have two lists of POS tags combinations we can compare
# We need to get the sentences back from the tagging exercise and run some stats

# Create a list of original sentences from the ADJECTIVE phrase subset:
adjph_whole_sentences2 = []

# loop over the sentences in the adjective phrase sentences we created:
for sents in grammar_adjph_tags2:
    temp=''
    for (word, tag) in sents:
        temp += word+' '
        adjph_whole_sentences2.append(temp)

print('\n', '=' * 50, '\n', len(adjph_whole_sentences2))

```

For for the second text I selected was "melville-moby_dick.txt"

Text2 - Top adjective phrases by frequency:

```

-----
so much                23
so many                21
once more              19
not so much            10
too much               9
as much                8
so wide                6
very large             5
so good                4
very little            4
as good                4
so saying              4
little more            4
not much               4
so small               4
so full                4
most appalling         4
so nigh                4
very curious           4
very much              4
so great               3
so important           3
still stranger          3
very similar           3
still more curious     3
very old               3
Very good              3
so remarkable          3
well nigh              3
still greater          3
so strange             3

```

very easy	3
however much	3
far other	3
very learned	3
very severe	3
not true	2
most monstrous	2
very great	2
so little	2
most convenient	2
most dangerous	2
just previous	2
too late	2
more wonderful	2
still further	2
most direful	2
most violent	2
somewhat similar	2
most conscientious	2
Length of adjective phrase sentences:	1627

Text2 - Top adverb phrases by frequency:

not only	50
as well	35
not so	28
so long	18
so much	15
so far	13
not yet	13
thus far	12
so soon	10
ere long	10
so very	9
as soon	8
even so	8
back again	8
so often	7
very much	6
even then	6
down there	6
never yet	6
not altogether	5
far away	5
Well then	5
not very	5
so strangely	5
just now	5
almost entirely	5
not always	5
still further	4
not then	4
even now	4
not possibly	4
not well	4
ye now	4
then again	4

very well	4
yet not	4
there again	4
not even	4
almost wholly	4
not wholly	4
aloft there	4
ye not	4
not now	4
very often	4
as much	3
forward again	3
not far	3
pretty nearly	3
very probably	3
just here	3
Length of adverb phrase sentences:	1235

Text2 - Top 50 Adjective Tokens

old	429
other	409
great	290
last	275
such	258
more	248
little	238
same	210
own	201
long	188
good	173
first	172
many	161
white	158
much	131
small	121
whole	115
full	111
poor	104
few	92
thy	92
certain	87
strange	87
high	86
most	84
dead	80
black	79
large	76
wild	76
least	75
young	75
true	73
vast	69
sperm	67
general	64
present	64
whale	61

hard	60
thou	60
entire	55
fine	54
curious	53
open	52
second	52
short	51
broad	51
lower	51
best	50
particular	49
ye	48

Text2 - Top 50 Adverb Tokens

not	1170
so	769
now	645
then	570
only	324
there	321
still	299
very	294
here	270
yet	263
more	252
again	252
n't	250
never	195
ever	190
most	190
almost	186
too	182
even	179
far	155
away	151
well	148
once	142
So	141
Now	139
down	138
as	119
long	114
soon	113
just	110
thus	107
also	88
much	87
back	87
perhaps	82
indeed	82
up	81
however	80
always	80
sometimes	76
rather	68

often	68
together	64
enough	59
Then	55
ere	54
ye	51
Well	51
aloft	49
therefore	49

Text2 - Top 50 Noun Tokens

whale	718
Ahab	495
man	472
ship	441
sea	349
time	316
boat	280
Whale	270
head	266
way	262
Stubb	251
Queequeg	244
whales	232
men	231
Captain	212
hand	194
Starbuck	191
thing	188
side	178
ye	172
world	166
water	166
Pequod	164
day	157
deck	157
eyes	155
sort	151
CHAPTER	150
part	148
boats	140
air	138
life	135
crew	135
Sperm	133
things	132
round	131
night	130
God	128
feet	125
hands	120
something	119
thou	113
body	110
sir	110
times	110

line	108
captain	108
moment	105
place	105
Flask	104

Text2 - Top 50 Verb Tokens

is	1722
was	1639
be	1027
had	767
have	763
were	679
are	586
's	426
been	415
do	363
said	302
has	291
seemed	282
did	264
say	233
see	229
being	219
go	178
made	175
seen	160
know	147
come	138
cried	138
came	130
take	117
thought	114
found	113
tell	113
called	112
saw	109
let	108
think	107
make	105
ye	98
went	97
heard	96
does	91
get	90
stood	86
am	84
seems	84
seem	84
going	79
look	76
done	74
known	73
put	70
give	68
round	67

stand 66

```

=====
3487
In [8]: #-----

In [9]: # PART 1.3a: PERFORM STATISTICS:
# Now we have two lists of POS tags combinations we can compare
# We need to get the sentences back from the tagging exercise and run some stats
print('For for the first text I selected was "' + fileName1)

# CREATE LIST OF ORIGINAL SENTENCES FROM THE ADJECTIVE PHRASE SUBSET:
adjph_whole_sentences1 = []
for sent in grammar_adjph_tags1:
    temp = ''
    for (word, tag) in sent:
        temp += word + ' '
    adjph_whole_sentences1.append(temp)
print('\nText1 - Length of Adjective Phrase Whole Sentences:\n', '-'*75)
print(len(adjph_whole_sentences1))

# CREATE LIST OF ORIGINAL SENTENCES FROM THE ADVERB PHRASE SUBSET:
advph_whole_sentences1 = []
for sent in grammar_advph_tags1:
    temp = ''
    for (word, tag) in sent:
        temp += word + ' '
    advph_whole_sentences1.append(temp)
print('\nText1 - Length of Adverb Phrase Whole Sentences:\n', '-'*75)
print(len(advph_whole_sentences1))

# COBINE LISTS TOGETHER TO HAVE A SINGLE LIST OF ADJECTIVE/ADVERB PHRASES:
# Useful to know which sentences are heavy in qualifiers
adv_adj_phrase_sentences1 = adjph_whole_sentences1
for sent in advph_whole_sentences1:
    # if a sentence is not in the adjective phrases list imported
    if sent not in adv_adj_phrase_sentences1:
        # attach that sentence
        adv_adj_phrase_sentences1.append(sent)
print('\nText1 - Length of Adverb + Adjective Phrase Whole Sentences:\n', '-'*75)
print(len(adv_adj_phrase_sentences1))

#####

# Following our NLTK textbook, Writing Structural Programs chapter
# section on Procedural vs Declarative style (http://www.nltk.org/book\_1ed/ch04.html)
#####

# CORPUS STATISTICS--SENTENCES LENGTH
# Calculating the average length of sentences in the entire corpus
# from http://www.nltk.org/book\_1ed/ch04.html
total_corpus1 = sum(len(sent) for sent in textsplitted) # remember: 'textsplitted' is our
text split into sentences
print('\nText1 - Average Length of Sentences:\n', '-'*75)

```

```
print(total_corpus1 / len(textsplit1))
```

```
# CALCULATE THE AVERAGE LENGTH OF ADJECTIVE PHRASE SENTENCE:
# We can then compare the average length of the adjective phrases to
# the average sentences we calculated for all sentences in the corpus
total_adjph_sentences1 = sum(len(sent) for sent in adjph_whole_sentences1)
# adjph_whole_sentences stores our adjective phrases
print('\nText1 - Average Length of Adjective Phrases:\n', '-'*75)
print(total_adjph_sentences1 / len(adjph_whole_sentences1))
```

For for the first text I selected was "carroll-alice.txt

Text1 - Length of Adjective Phrase Whole Sentences:

464

Text1 - Length of Adverbe Phrase Whole Sentences:

484

Text1 - Length of Adverbe + Adjective Phrase Whole Sentences:

633

Text1 - Average Length of Sentences:

87.26892307692307

Text1 - Average Length of Adjective Phrases:

9.262243285939968

In [10]: #-----

In [11]: # PART 1.3b: PERFORM STATISTICS:
Now we have two lists of POS tags combinations we can compare
We need to get the sentences back from the tagging exercise and run some stats
print('For for the second text I selected was "' + fileName2)

```
# CREATE LIST OF ORIGINAL SENTENCES FROM THE ADJECTIVE PHRASE SUBSET:
adjph_whole_sentences2 = []
for sents in grammar_adjph_tags2:
    temp=''
    for (word, tag) in sents:
        temp += word+' '
    adjph_whole_sentences2.append(temp)
print('\nText2 - Length of Adjective Phrase Whole Sentences:\n', '-'*75)
print(len(adjph_whole_sentences2))
```

```
# CREATE LIST OF ORIGINAL SENTENCES FROM THE ADVERB PHRASE SUBSET:
advph_whole_sentences2 = []
for sents in grammar_advph_tags2:
    temp = ''
    for (word, tag) in sents:
```

```

        temp += word+ ' '
        advph_whole_sentences2.append(temp)
print('\nText2 - Length of Adverbe Phrase Whole Sentences:\n', '-'*75)
print(len(advph_whole_sentences2))

# COBMINE LISTS TOGETHER TO HAVE A SINGLE LIST OF ADJECTIVE/ADVERB PHRASES:
# Useful to know which sentences are heavy in qualifiers
adv_adj_phrase_sentences2 = adjph_whole_sentences2
for sent in advph_whole_sentences2:
    # if a sentence is not in the adjective phrases list imported
    if sent not in adv_adj_phrase_sentences2:
        # attach that sentence
        adv_adj_phrase_sentences2.append(sent)
print('\nText2 - Length of Adverbe + Adjective Phrase Whole Sentences:\n', '-'*75)
print(len(adv_adj_phrase_sentences2))

#####

# Following our NLTK textbook, Writing Structural Programs chapter
# section on Procedural vs Declarative style (http://www.nltk.org/book\_1ed/ch04.html)
#####

# CORPUS STATISTICS--SENTENCES LENGTH
# Calculating the average length of sentences in the entire corpus
# from http://www.nltk.org/book\_1ed/ch04.html
total_corpus2 = sum(len(sent) for sent in textsplitted) # remember: 'textsplitted' is our
text split into sentences
print('\nText2 - Average Length of Sentences:\n', '-'*75)
print(total_corpus2 / len(textsplitted))

# CALCULATE THE AVERAGE LENGTH OF ADJECTIVE PHRASE SENTENCE:
# We can then compare the average length of the adjective phrases to
# the average sentences we calculated for all sentences in the corpus
total_adjph_sentences2 = sum(len(sent) for sent in adjph_whole_sentences2)
# adjph_whole_sentences2 stores our adjective phrases
print('\nText2 - Average Length of Adjective Phrases:\n', '-'*75)
print(total_adjph_sentences2 / len(adjph_whole_sentences2))

```

For for the second text I selected was "melville-moby_dick.txt"

Text2 - Length of Adjective Phrase Whole Sentences:

3487

Text2 - Length of Adverbe Phrase Whole Sentences:

2563

Text2 - Length of Adverbe + Adjective Phrase Whole Sentences:

4234

Text2 - Average Length of Sentences:

123.65489240763297

Text2 - Average Length of Adjective Phrases:

10.43410486537553

PART 2: INTERPRETATION OF THE RESULTS:

- Provide interpretation of the Part 1 results.
- What do you learn from these results?

The sentiment analysis had mixed results. In terms of being able to distinguish between the two books (Alice in Wonderland and Moby Dick), the sentiment analysis did allow these books to be distinguished and identified. This is because the analysis did a good job at picking out specific sentiment phrases and did a great job with finding key nouns. The issue came in when trying to connect the specific sentiment to the specific objects or nouns with the two stories. This portion, in my opinion, is where the sentiment analysis faltered or stumbled. It was like we had a collection of objects and a collection of sentiments but not a good way to know which one went with which.

A good analogy of the current state of sentiment analysis would be two tables in a database. One table has the sentiment phrases and the second had the objects or nouns. I believe the sentiment analysis techniques used can successfully populate those two tables. The part that is still missing and needs further development is relating those two tables together correctly. That is, identifying which records in one table are associated with the records in the other. Without that proper relating the sentiment analysis loses the context and thus becomes difficult to interpret. The good news is that since we know what the 'missing link' is (pun intended) we know where we need to spend our efforts.

I learned that NLP is much more complicated than I ever gave thought to in the past and that substantial progress has been made in this area. However, there is still much more to do to get to a point in which a computer can understand Natural Language. The challenges have been identified and while progress is slow it is definitely something that we will figure out one day. I'm not sure we will figure it out in 5 years or 50 years from now but I strongly believe it is something humanity will accomplish eventually.

On very macro scale I have come to believe that AI and NLP are far more related than I ever thought possible. While many metrics are undoubtedly available to give an indication of how advanced an AI system is I suspect in the end the truest test will be can we talk to it and it understand what we are saying. That will be the point in which AI will have reached a turning point in the development of these systems.

In [12]: #-----

PART 3: VISION FOR THE FUTURE

- The last thing to do is to envision how you would conduct a sentiment analysis on the review texts.

I can envision sentiment analysis being used in two complementary ways. In the short term, I think the sentiment analysis will be used to process large volumes of short texts in which the topic is highly focused. For example, these techniques could be used to analyze customer product reviews or comment boards. The reason for these limitations is due to the challenge of accurately and consistently attributing specific sentiment to a specific topic or subject.

In addition, in the near term, I believe there will be a great deal to improve our systems ability to accurately contribute sentiment to a specific subject or topic. This will be a major topic of study, analysis, and will surely have triumphs. Unfortunately it is highly likely that progress will remain steady and slow. This gives a great deal of opportunity for research and development for those interested in this field of exploration.

In the long term, I can see sentiment analysis being used in virtually every area of our lives. As sentiment analysis improves and accuracy increases the application space will grow wider and more complicated. So, in the short term, perhaps relatively simple texts in which the subject or topic is limited. Over time more complicated texts and stories until reading is as easy for the computer to do as doing mathematics currently is. This is truly an exciting time in the field.

In [13]: #-----

In []:

In []: