

Link:- <https://github.com/shail10/NLP-project>

Language Learners

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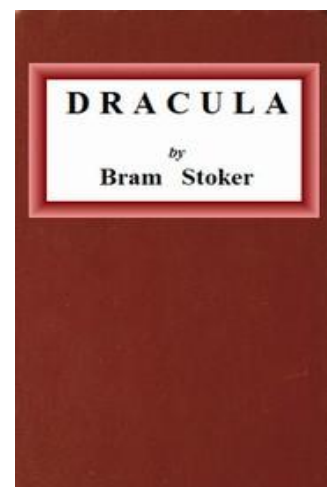
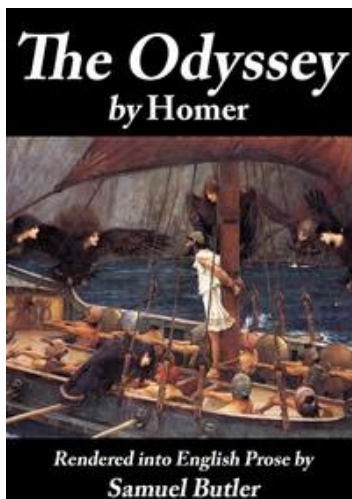
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NLP PROJECT - ROUND 1

OVERVIEW

In this project, we perform text analysis on two of our chosen books from Gutenberg. “Dracula” and “Odyssey.” After that, we will apply POS Tagging on both books.

IMPORTED BOOKS



OBJECTIVES

- Import the text, let's call it T1 and T2
- Perform simple text preprocessing steps and tokenising the text T1 and T2.
- Analyze the frequency distribution of tokens in T1 and T2 separately.
- Create a Word Cloud of T1 and T2 using the token that we have got.
- Remove the stop words from T1 and T2 and then again create a word cloud. Comparison with word cloud before the removal of stop words.
- Evaluate the relationship between the word length and frequency for both T1 and T2.
- Do PoS Tagging for both T1 and T2 using any tokenising of the four tag sets studied in the class and get the distribution of various tags.

Libraries Used:-

```
[ ] #imports
import nltk
from nltk.tokenize import word_tokenize #for tokenizing the words
from nltk.stem import PorterStemmer #For potter stemming
from nltk.stem import WordNetLemmatizer #for lemmatization/to get lemma of the word
from nltk.corpus import stopwords      #To remove stop words
from urllib.request import urlopen

import inflect
import re #for the purpose of regular expressions
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
from wordcloud import WordCloud
```

```
➤ nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
True
```

```
[ ] nltk.download('punkt')
```

```
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]   Package punkt is already up-to-date!
True
```

```
[ ] nltk.download('wordnet')
```

```
[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data]   Package wordnet is already up-to-date!
True
```

```
[ ] p = inflect.engine()
    ps = PorterStemmer()
    lemmatizer = WordNetLemmatizer()
```

Data preprocessing

1) Importing the text using the urllib

```
[ ] #URLs for Books
url1 = 'https://www.gutenberg.org/cache/epub/1727/pg1727.txt' #Odyssey
url2 = 'https://www.gutenberg.org/files/345/345-0.txt' #Dracula
```

```
[ ] #first we will read both the books and print few initial lines
T1_odyssey = urlopen(url1).read()
T2_dracula = urlopen(url2).read()
```

```
[ ] T1_odyssey
```

```
b'\xef\xbb\xbfThe Project Gutenberg eBook of The Odyssey, by Homer\r\n\r\nThis eBook is for the use of anyone anywhere in the United States and\r\nmost other parts of the world at no cost and with almost no
< ▶
```

```
[ ] T2_dracula
```

```
b'\xef\xbb\xbfThe Project Gutenberg eBook of Dracula, by Bram Stoker\r\n\r\nThis eBook is for the use of anyone anywhere in the United States and\r\nmost other parts of the world at no cost and with almost
< ▶
```

```
[ ] T1_odyssey = T1_odyssey.decode('utf-8')
```

```
[ ] T2_dracula = T2_dracula.decode('utf-8')
```

Performing simple text-preprocessing steps and tokenizing both T1_odeyssey and T2_dracula

We are removing all the unnecessary text from the files.

```
[ ] def discard_from_odyssey(text):
    sid = text.find('THE ODYSSEY')
    eid = text.find('*** END OF THE PROJECT GUTENBERG EBOOK THE ODYSSEY ***)
    print("Discarding Before - ", sid)
    print("Discarding After - ", eid)
    text = text[sid:eid]
    return text
```

```
[ ] def discard_from_dracula(text):
    sidx = text.find('DRACULA')
    eidx = text.find('*** END OF THE PROJECT GUTENBERG EBOOK DRACULA ***')
    print("Discarding Before - ", sidx)
    print("Discarding After - ", eidx)
    text = text[sidx:eidx]
    return text
```

```
[ ] T1_odyssey = discard_from_odyssey(T1_odyssey)
```

Discarding Before - 772
Discarding After - 691478

```
[ ] T2_dracula = discard_from_dracula(T2_dracula)
```

Discarding Before - 805
Discarding After - 862449

[illegible][illegible]

```
[ ] T1_odyssey = T1_odyssey.lower()
    T2_dracula = T2_dracula.lower()
```

Using regular expression to decontract certain words to standard form for better text understanding

```
def transforming(text):
    #removing URL
    text = re.sub(r"http[s]+", "", text)

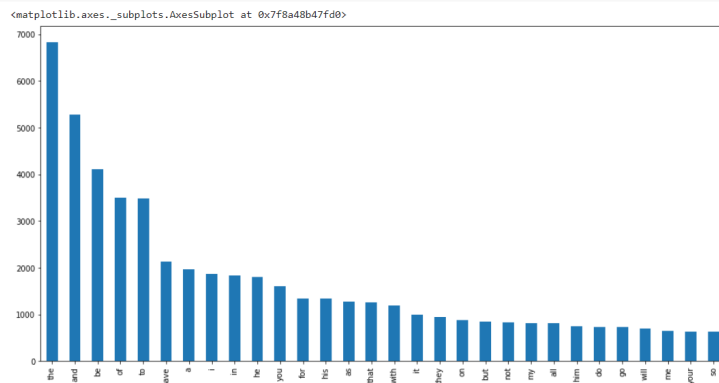
    #Decontracting most common words
    text = re.sub(r"couldn't", "could not", text)
    text = re.sub(r"aren't", "are not", text)
    text = re.sub(r"won't", "will not", text)
    text = re.sub(r"can't", "can not", text)
    text = re.sub(r"n't", "not", text)
    text = re.sub(r"re'", "are", text)
    text = re.sub(r"'s", "is", text)
    text = re.sub(r"'d", "would", text)
    text = re.sub(r"'ll", "will", text)
    text = re.sub(r"'t", "not", text)
    text = re.sub(r"'ve", "have", text)
    text = re.sub(r"'m", "am", text)

    return text
```

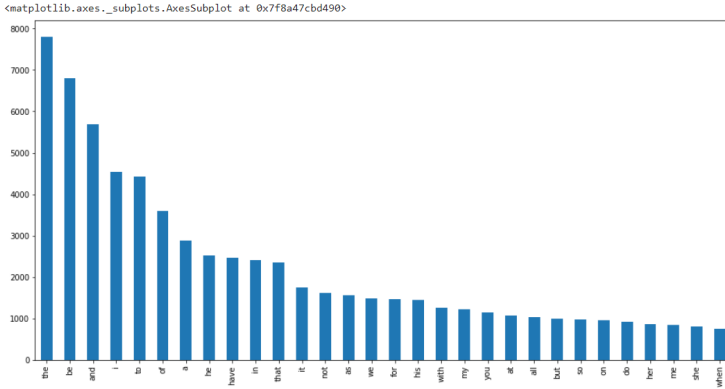
```
[ ] T1_odyssey = transforming(T1_odyssey)
    T2_dracula = transforming(T2_dracula)
```

Tokenizing both the book

dracula *** dracula dracula _by_ bram stoker [illustration: colophon] new york grosset & dunlap publishers copyright, 1897, in the united states of america, according to act of congress, by bram stoker. [all rights reserved.] printed in the united states at the country life press, garden city, n.y. to my dear friend hommy-beg contents chapter i. jonathan harker's journal chapter ii. jonathan harker's journal chapter iii. jonathan harker's journal chapter iv. jonathan harker's journal chapter v. letters-lucy and mina chapter vi. mina murray's journal chapter vii. cutting from "the dailygraph," eight august t chapter viii. mina murray's journal chapter ix. mina murray's journal chapter x. mina murray's journal chapter xi. lucy westena's diary chapter xii. dr. seward's diary chapter xiii. dr. seward's diary chapter xiv. mina harker's journal chapter xv. dr. seward's diary chapter xvi. dr. seward's diary chapter xvii. dr. seward's diary chapter xviii. dr. seward's diary chapter xix. jona.

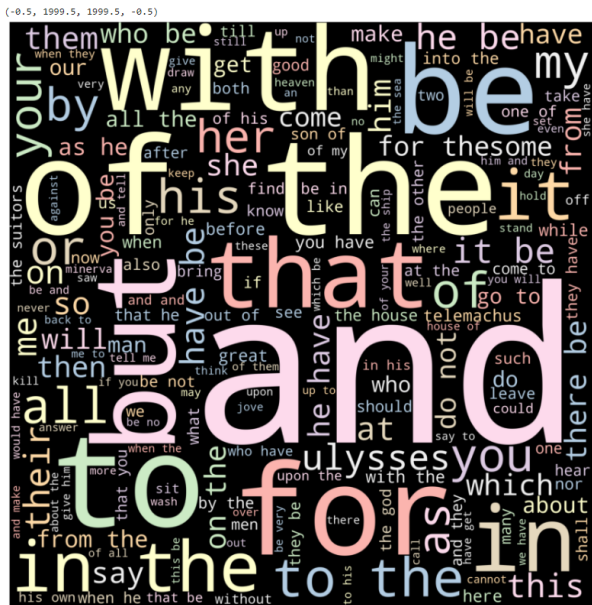


```
[ ] plt.figure(figsize=(16,8))
    tokens = word_tokenize(T2_dracula)
    pd.Series(tokens).value_counts()[:30].plot(kind='bar')
```



Generating Word Cloud

```
plt.figure(figsize=(15,15))
wordcloud = WordCloud(width = 2000, height = 2000, background_color = 'black', stopwords = [], colormap='Pastell').generate(T1_odyssey)
plt.imshow(wordcloud)
plt.axis('off')
```





From the above word cloud visualisation, we can infer that the most frequently are mainly stopped words like 'to', 'of', 'be', 'the'. These words do not contribute to the meaning of the sentence. These stop words can easily be removed and the resulting in a better word cloud.

Removing the STOPWORDS and again generating Word Cloud to identify the potential differences between the word clouds before and after removing the STOPWORDS.

```

1 stop_words = set(stopwords.words('english'))
def remove_stopwords(text):
    tokens = word_tokenize(text)
    tokens = [words for words in tokens if words not in stop_words]
    return ' '.join(tokens)

[ ] T1_odyssey_ = remove_stopwords(T1_odyssey)
    T2_dracula_ = remove_stopwords(T2_dracula)

[ ] plt.figure(figsize=(15,15))
    wordcloud = WordCloud(width = 2000, height = 2000, background_color = 'black', stopwords = [], colormap="Pastell").generate(T1_odyssey_)
    plt.imshow(wordcloud)
    plt.axis('off')

```

[illegible]

window ask 1999.5, 1999.5, -0.5) takes us shall two
 fear drive dear life answer pass wish seaward sort two
 god strange mind another wake ever too heart follow wind meet case bring
 harker whole door call rest mina much back grow placemight si
 without mean great friend arthur behind far hour frequency
 begin body first felt find sleep lay leave since
 want love eye speak draw use upon van poor three away send husband
 lie white throw fell live lose
 think hold try morning every poor
 wou d know
 count long professor van helsing never watch even
 full set wait close point
 all house men raise help open something let night work
 get many man yet put turn light Jonathan hand hear blood Lucy quite end
 whilst suppose all
 thing though well part die right need till show old
 must time could see may like make
 become move always letter read young box sosh head woman nothing touch
 time could see may like make
 could seaword other stalk start horse feel little write

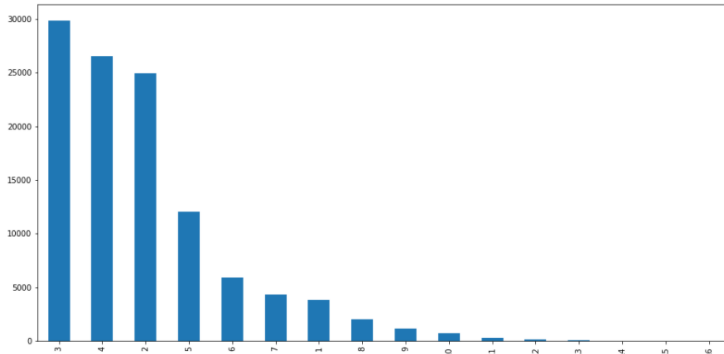
The words with no meaning are removed from the text after removing all the stop words.
Now more relevant words can be displayed with word clouds with words like 'would', 'know', 'come', 'take'.

Evaluating the relationship between the word length and frequency for both T1_odyssey and T2_dracula

Before removing StopWords

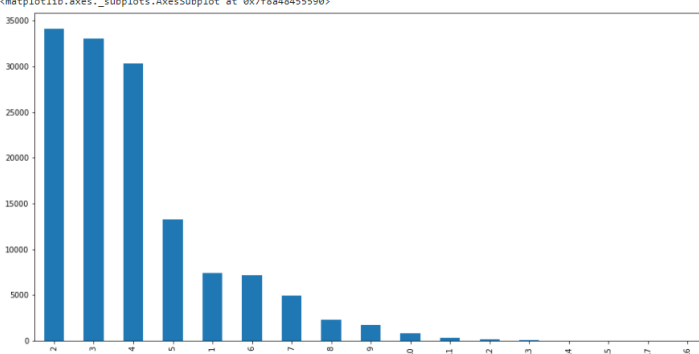
```
# For T1_odyssey
plt.figure(figsize=(16,8))
tokens = word_tokenize(T1_odyssey)
length = [len(words) for words in tokens]
pd.Series(length).value_counts()[1:30].plot(kind='bar')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8a468f9e10>



```
# For T2_dracula
plt.figure(figsize=(16,8))
tokens = word_tokenize(T2_dracula)
length = [len(words) for words in tokens]
pd.Series(length).value_counts()[1:30].plot(kind='bar')
```

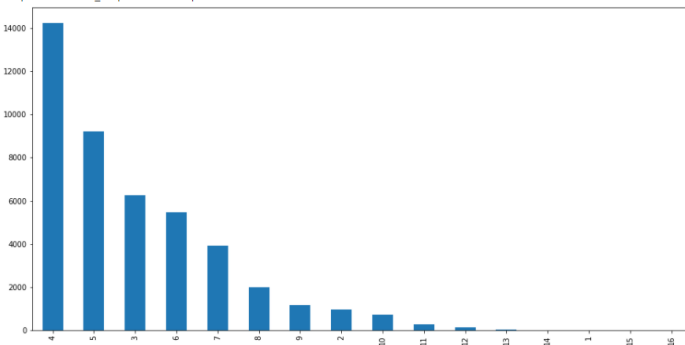
<matplotlib.axes._subplots.AxesSubplot at 0x7f8a48455990>



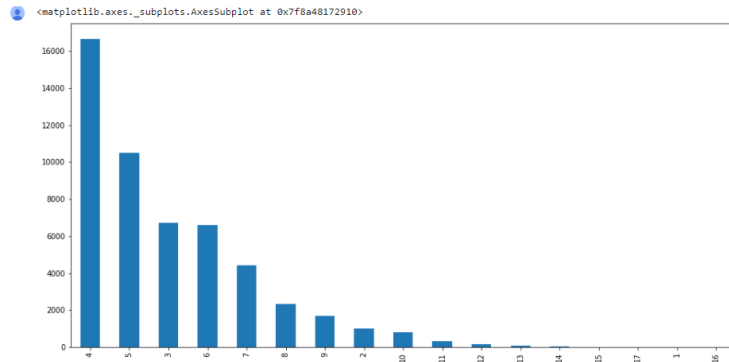
After removing StopWords

```
# For T1_odyssey
plt.figure(figsize=(16,8))
tokens = word_tokenize(T1_odyssey_)
length = [len(words) for words in tokens]
pd.Series(length).value_counts()[1:30].plot(kind='bar')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8a474f0310>



```
# For T2_dracula
plt.figure(figsize=(16,8))
tokens = word_tokenize(T2_dracula_)
length = [len(words) for words in tokens]
pd.Series(length).value_counts()[1:30].plot(kind='bar')
```



The number of words of the length of 2 and 3 has been decreased after removing stopwords. This is since stopping words like 'be', 'of' have been removed. Apart from that, new comments have been emerging, the stop words of length 3,4,5.

POS Tagging

```
[ ] from collections import Counter

[ ] def tag_treebank(text):
    tokenized=nlk.word_tokenize(text)
    tagged=nlk.pos_tag(tokenized)
    return tagged

[ ] def get_counts(tags):
    counts = Counter( tag for word, tag in tags)
    return counts

[ ] nltk.download('averaged_perceptron_tagger')

[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data] /root/nltk_data...
[nltk_data] Unzipping taggers/averaged_perceptron_tagger.zip.
True

[ ] odessey_tags=tag_treebank(T1_odessey)
odessey_pos_count=get_counts(odessey_tags)
dracula_tags=tag_treebank(T2_dracula)
dracula_pos_count=get_counts(dracula_tags)
```

```
[ ] len(odessey_pos_count)
```

```
32
```

```
[ ] odessey_pos_count
```

```
Counter({'S': 2,  
        'CC': 6782,  
        'CD': 774,  
        'DT': 11036,  
        'EX': 189,  
        'FW': 31,  
        'IN': 16283,  
        'JJ': 7104,  
        'JJR': 297,  
        'JJS': 201,  
        'MD': 2551,  
        'NN': 20858,  
        'NNP': 50,  
        'NNS': 1344,  
        'PDT': 248,  
        'PRP': 8903,  
        'PPRS': 2649,  
        'RB': 5066,  
        'RBR': 93,  
        'RBS': 65,  
        'RP': 668,  
        'TO': 3479,  
        'VB': 12382,  
        'VBD': 755,  
        'VBG': 204,  
        'VBN': 2292,  
        'VBP': 5173,  
        'VBZ': 621,  
        'MDT': 653,  
        'WP': 959,  
        'WPS': 43,  
        'WRB': 981})
```

```
[ ] len(dracula_pos_count)
```

```
33
```

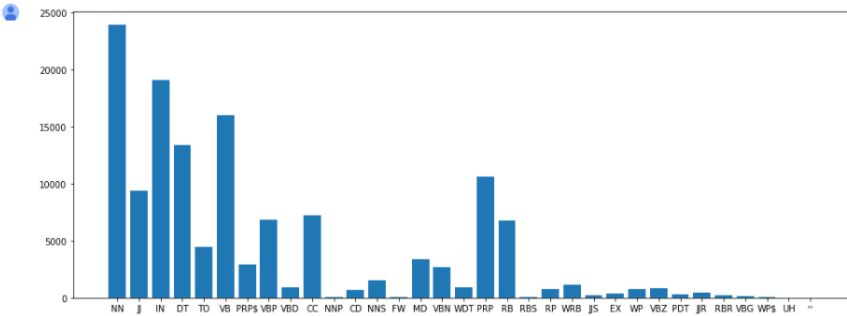
```
[ ] dracula_pos_count
```

```
Counter('': 1,  
        'CC': 7210,  
        'CD': 650,  
        'DT': 13401,  
        'EX': 362,  
        'FW': 50,  
        'IN': 19083,  
        'JJ': 9373,  
        'JJR': 426,  
        'JJS': 210,  
        'MD': 3394,  
        'NN': 23928,  
        'NNP': 26,  
        'NNS': 1504,  
        'PDT': 310,  
        'PRP': 10607,  
        'PPRS': 2915,  
        'RB': 6733,  
        'RBR': 192,  
        'RBS': 47,  
        'RP': 776,  
        'TO': 4429,  
        'UH': 2,  
        'VB': 16014,  
        'VBD': 886,  
        'VBG': 102,  
        'VBN': 2685,  
        'VBP': 6873,  
        'VBZ': 795,  
        'MDT': 894,  
        'WP': 719,  
        'WPS': 30,  
        'WRB': 1133})
```

```
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(16,6))  
plt.bar(range(len(dracula_pos_count)), list(dracula_pos_count.values()), align='center')  
plt.xticks(range(len(dracula_pos_count)), list(dracula_pos_count.keys()))
```

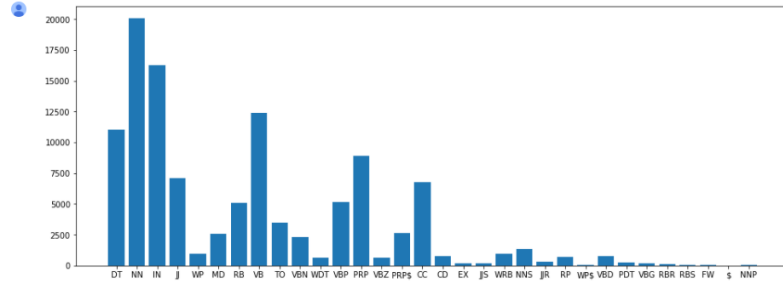
```
plt.show()
```



```
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(16,6))  
plt.bar(range(len(odessey_pos_count)), list(odessey_pos_count.values()), align='center')  
plt.xticks(range(len(odessey_pos_count)), list(odessey_pos_count.keys()))
```

```
plt.show()
```



From this we can infer that the highest occurring tag is 'NN', and 'Determinant' Tags are on the lower frequency side. This is largely due to the removal of stopwords before POS Tagging.

CONCLUSION:-

We have learnt how to perform Text Processing, Tokenization and answer everything we asked of us using proper imports and libraries like pandas, matplotlib, NLTK.