Forward School

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Program Name: FRONT-END SOFTWARE DEVELOPMENT

Title : Basic Text Analytics using NLTK

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Introduction : learning how to do nltk

Conclusion : learned how to use nltk

NLTK - Basic Text Analytics

- Word & Sentence tokenizer
- Parts of Speech (POS) tagger
- Extracting Entities

Installation

conda install -c conda-forge nltk

In [41]:
```
import nltk

nltk.download('punkt')
nltk.download('stopwords')
nltk.download("wordnet")
nltk.download('averaged_perceptron_tagger')
nltk.download('brown')
nltk.download('maxent_ne_chunker')
nltk.download('words')
nltk.download('names')
nltk.download('movie_reviews')
```

```
[nltk_data] Downloading package punkt to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package punkt is already up-to-date!
[nltk_data] Downloading package stopwords to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
[nltk_data] Downloading package wordnet to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package wordnet is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package averaged_perceptron_tagger is already up-to-
[nltk_data]       date!
[nltk_data] Downloading package brown to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package brown is already up-to-date!
[nltk_data] Downloading package maxent_ne_chunker to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package maxent_ne_chunker is already up-to-date!
[nltk_data] Downloading package words to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package words is already up-to-date!
[nltk_data] Downloading package names to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package names is already up-to-date!
[nltk_data] Downloading package movie_reviews to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Unzipping corpora\movie_reviews.zip.
```

Out[41]: True

Text Analytics for Beginners using NLTK

Learn How to analyze text using NLTK. Analyze people's sentiments and classify movie reviews.

In today's area of internet and online services, data is generating at incredible speed and amount. Generally, Data analyst, engineer, and scientists are handling relational or tabular data. These tabular data columns have either numerical or categorical data. Generated data has a variety of structures such as text, image, audio, and video. Online activities such as articles, website text, blog posts, social media posts are generating unstructured textual data. Corporate

and business need to analyze textual data to understand customer activities, opinion, and feedback to successfully derive their business. To compete with big textual data, text analytics is evolving at a faster rate than ever before.

Text Analytics has lots of applications in today's online world. By analyzing tweets on Twitter, we can find trending news and peoples reaction on a particular event. Amazon can understand user feedback or review on the specific product. BookMyShow can discover people's opinion about the movie. Youtube can also analyze and understand peoples viewpoints on a video.

In this tutorial, you are going to cover the following topics:

Text Analytics and NLP Compare Text Analytics, NLP and Text Mining Text Analysis Operations using NLTK Tokenization Stopwords Lexicon Normalization such as Stemming and

Text Analytics and NLP

Text communication is one of the most popular forms of day to day conversion. We chat, message, tweet, share status, email, write blogs, share opinion and feedback in our daily routine. All of these activities are generating text in a significant amount, which is unstructured in nature. I this area of the online marketplace and social media, It is essential to analyze vast quantities of data, to understand peoples opinion.

NLP enables the computer to interact with humans in a natural manner. It helps the computer to understand the human language and derive meaning from it. NLP is applicable in several problematic from speech recognition, language translation, classifying documents to information extraction. Analyzing movie review is one of the classic examples to demonstrate a simple NLP Bag-of-words model, on movie reviews.

Compare Text Analytics, NLP and Text Mining

Text mining also referred to as text analytics. Text mining is a process of exploring sizeable textual data and find patterns. Text Mining process the text itself, while NLP process with the underlying metadata. Finding frequency counts of words, length of the sentence, presence/absence of specific words is known as text mining. Natural language processing is one of the components of text mining. NLP helps identified sentiment, finding entities in the sentence, and category of blog/article. Text mining is preprocessed data for text analytics. In Text Analytics, statistical and machine learning algorithm used to classify information.

Text Analysis Operations using NLTK

NLTK is a powerful Python package that provides a set of diverse natural languages algorithms. It is free, opensource, easy to use, large community, and well documented. NLTK consists of the most common algorithms such as tokenizing, part-of-speech tagging, stemming, sentiment analysis, topic segmentation, and named entity recognition. NLTK helps the computer to analysis, preprocess, and understand the written text.

Tokenization

Tokenization is the first step in text analytics. The process of breaking down a text paragraph into smaller chunks such as words or sentence is called Tokenization. Token is a single entity that is building blocks for sentence or paragraph.

Sentence Tokenization

Sentence tokenizer breaks text paragraph into sentences.

In [10]:
```
import nltk

from nltk.tokenize import sent_tokenize, word_tokenize

# conda install -c anaconda nltk

url = 'https://www.timeout.com/kuala-lumpur/food-and-drink/food-reviews'

text = '''Table & Apron – formerly The Kitchen Table Restaurant & Bakery – does
'''

sentences = nltk.sent_tokenize(text)

print(sentences)
```

```
['Table & Apron – formerly The Kitchen Table Restaurant & Bakery – doesn’t ex
ist to disrupt the scene.', 'From the outside, it barely stretches the bounda
ries of what is an already saturated restaurant-cum-bakery scene.', 'But none
of it matters.', 'Because right from its birth in 2014, Table & Apron has pro
ven to be a restaurant that has in spades a component so elementary yet so ra
re – heart.', 'Through hard work, dedication and all the boring old-fashioned
virtues of an honest operation, owner Marcus Low and his team have carved for
us a little treasure in Damansara Kim.', '(Credit must also be given to forme
r co-owner Mei Wan Tan.)', 'The narcissism you’ll find in so many KL']
```

Here, the given text is tokenized into sentences.

Word Tokenization

Word tokenizer breaks text paragraph into words.

In [11]:
```
words = word_tokenize(text)

print(words)
```

```
['Table', '&', 'Apron', '–', 'formerly', 'The', 'Kitchen', 'Table', 'Restaura
nt', '&', 'Bakery', '–', 'doesn', '’', 't', 'exist', 'to', 'disrupt', 'the',
'scene', '.', 'From', 'the', 'outside', ',', 'it', 'barely', 'stretches', 'th
e', 'boundaries', 'of', 'what', 'is', 'an', 'already', 'saturated', 'restaura
nt-cum-bakery', 'scene', '.', 'But', 'none', 'of', 'it', 'matters', '.', 'Bec
ause', 'right', 'from', 'its', 'birth', 'in', '2014', ',', 'Table', '&', 'Apr
on', 'has', 'proven', 'to', 'be', 'a', 'restaurant', 'that', 'has', 'in', 'sp
ades', 'a', 'component', 'so', 'elementary', 'yet', 'so', 'rare', '–', 'hear
t', '.', 'Through', 'hard', 'work', ',', 'dedication', 'and', 'all', 'the',
'boring', 'old-fashioned', 'virtues', 'of', 'an', 'honest', 'operation', ',',
'owner', 'Marcus', 'Low', 'and', 'his', 'team', 'have', 'carved', 'for', 'u
s', 'a', 'little', 'treasure', 'in', 'Damansara', 'Kim', '.', '(', 'Credit',
'must', 'also', 'be', 'given', 'to', 'former', 'co-owner', 'Mei', 'Wan', 'Ta
n', '.', ')', 'The', 'narcissism', 'you', '’', 'll', 'find', 'in', 'so', 'man
y', 'KL']
```

In [12]:
```
# depending on domain, tokenizer
print(nltk.wordpunct_tokenize(text))
```

```
['Table', '&', 'Apron', '–', 'formerly', 'The', 'Kitchen', 'Table', 'Restaura
nt', '&', 'Bakery', '–', 'doesn', '’', 't', 'exist', 'to', 'disrupt', 'the',
'scene', '.', 'From', 'the', 'outside', ',', 'it', 'barely', 'stretches', 'th
e', 'boundaries', 'of', 'what', 'is', 'an', 'already', 'saturated', 'restaura
nt', '-', 'cum', '-', 'bakery', 'scene', '.', 'But', 'none', 'of', 'it', 'mat
ters', '.', 'Because', 'right', 'from', 'its', 'birth', 'in', '2014', ',', 'T
able', '&', 'Apron', 'has', 'proven', 'to', 'be', 'a', 'restaurant', 'that',
'has', 'in', 'spades', 'a', 'component', 'so', 'elementary', 'yet', 'so', 'ra
re', '–', 'heart', '.', 'Through', 'hard', 'work', ',', 'dedication', 'and',
'all', 'the', 'boring', 'old', '-', 'fashioned', 'virtues', 'of', 'an', 'hone
st', 'operation', ',', 'owner', 'Marcus', 'Low', 'and', 'his', 'team', 'hav
e', 'carved', 'for', 'us', 'a', 'little', 'treasure', 'in', 'Damansara', 'Ki
m', '.', '(', 'Credit', 'must', 'also', 'be', 'given', 'to', 'former', 'co',
'-', 'owner', 'Mei', 'Wan', 'Tan', '.)', 'The', 'narcissism', 'you', '’', 'l
l', 'find', 'in', 'so', 'many', 'KL']
```

Difference between word_tokenize and wordpunct_tokenize

Reference: [https://stackoverflow.com/questions/50240029/nltk-wordpunct-tokenize-vs-word-tokenize (https://stackoverflow.com/questions/50240029/nltk-wordpunct-tokenize-vs-word-tokenize)](https://stackoverflow.com/questions/50240029/nltk-wordpunct-tokenize-vs-word-tokenize)

Frequency Distribution

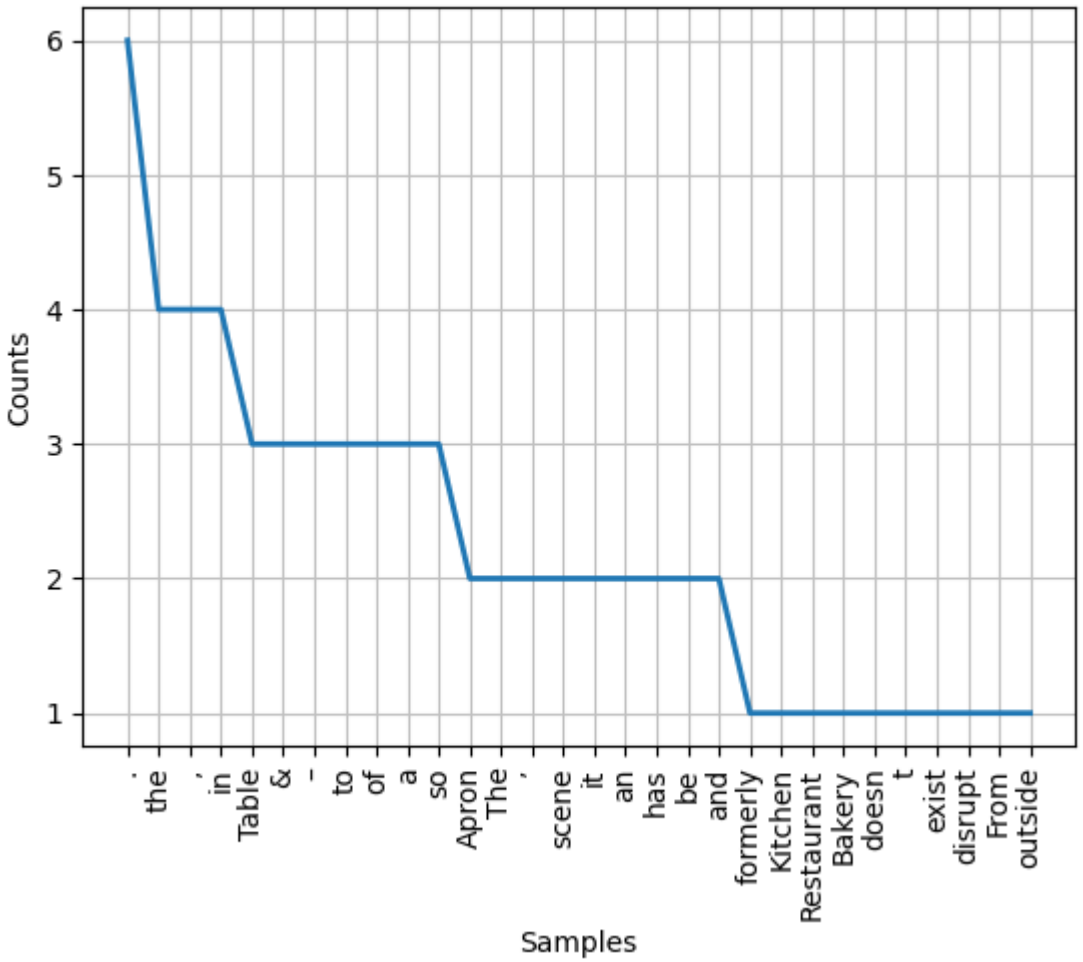
In [13]:
```
from nltk.probability import FreqDist

fdist= FreqDist(words)

print(fdist)
```

```
<FreqDist with 96 samples and 133 outcomes>
```

In [14]:
```
fdist.most_common(3)
```

Out[14]:
```
[('.', 6), ('the', 4), (',', 4)]
```

In [15]:
```
# Frequency Distribution Plot
import matplotlib.pyplot as plt
fdist.plot(30,cumulative=False)

plt.show()
```



Stopwords

Stopwords considered as noise in the text. Text may contain stop words such as is, am, are, this, a, an, the, etc.

In NLTK for removing stopwords, you need to create a list of stopwords and filter out your list of tokens from these words.

In [16]:
```
from nltk.corpus import stopwords
stop_words=set(stopwords.words("english"))
print(stop_words)
```

```
{'in', 'ours', 'couldn', "should've", 'few', 've', 'ma', 're', 'y', "does
n't", 'i', 'needn', 'had', 'd', 'which', "wasn't", 'nor', 'over', 'any', 'our
selves', 'with', 'mustn', 'weren', 'other', 'than', 'until', 't', 'above', "a
ren't", "wouldn't", 'll', "you're", "you've", 'there', 'those', "needn't", 'f
urther', 'this', 'yours', 'just', 'it', 'does', 'what', "haven't", 'an', 'sh
e', 'a', 'these', 'here', "mustn't", 'if', 'them', 'down', 'didn', 'how', 'th
emselves', 'because', 'now', 'doesn', 'against', 'theirs', 'at', 'the', 'ou
t', 'was', "mightn't", 'more', 'him', 'me', 'we', 'herself', 'through', 'sam
e', 'yourself', 'shan', 'hers', 'yourselves', 'her', 'being', 'such', 'its',
"it's", 'off', "you'll", 'while', 'my', 'our', 'where', 'itself', 'be', 'is',
'should', 'during', 'has', 'between', 'are', 'did', 'but', 'under', "don't",
'having', 'of', "hadn't", 'their', 'doing', "shan't", 'been', 'hadn', 'm', 'w
hy', "you'd", 'his', 'he', 'again', "she's", 'myself', "didn't", 'can', 'o',
"won't", 'they', "that'll", 'up', 'for', "shouldn't", 'very', 'ain', 'too',
'were', "isn't", 'before', 'wasn', 'hasn', 'below', 'shouldn', 'to', 'some',
'do', 'each', 'both', 'have', 'on', 'into', 'will', 'haven', 'when', 'most',
'then', 'who', 'by', 'not', 'once', 'from', 'you', 'own', 'whom', 'no', 'an
d', "hasn't", 'himself', 'about', "couldn't", 'am', 'as', 's', 'after', 'al
l', 'won', "weren't", 'aren', 'only', 'mightn', 'so', 'your', 'don', 'isn',
'that', 'or', 'wouldn'}
```

Removing Stopwords from our paragraphs

In [17]:
```
tokenized_sent = words

filtered_sent=[]
for w in tokenized_sent:
    if w not in stop_words:
        filtered_sent.append(w)
print("Tokenized Sentence:",tokenized_sent)
print()
print("Filtered Sentence:",filtered_sent)
```

```
Tokenized Sentence: ['Table', '&', 'Apron', '–', 'formerly', 'The', 'Kitche
n', 'Table', 'Restaurant', '&', 'Bakery', '–', 'doesn', '’', 't', 'exist', 't
o', 'disrupt', 'the', 'scene', '.', 'From', 'the', 'outside', ',', 'it', 'bar
ely', 'stretches', 'the', 'boundaries', 'of', 'what', 'is', 'an', 'already',
'saturated', 'restaurant-cum-bakery', 'scene', '.', 'But', 'none', 'of', 'i
t', 'matters', '.', 'Because', 'right', 'from', 'its', 'birth', 'in', '2014',
',', 'Table', '&', 'Apron', 'has', 'proven', 'to', 'be', 'a', 'restaurant',
'that', 'has', 'in', 'spades', 'a', 'component', 'so', 'elementary', 'yet',
'so', 'rare', '–', 'heart', '.', 'Through', 'hard', 'work', ',', 'dedicatio
n', 'and', 'all', 'the', 'boring', 'old-fashioned', 'virtues', 'of', 'an', 'h
onest', 'operation', ',', 'owner', 'Marcus', 'Low', 'and', 'his', 'team', 'ha
ve', 'carved', 'for', 'us', 'a', 'little', 'treasure', 'in', 'Damansara', 'Ki
m', '.', '(', 'Credit', 'must', 'also', 'be', 'given', 'to', 'former', 'co-ow
ner', 'Mei', 'Wan', 'Tan', '.', ')', 'The', 'narcissism', 'you', '’', 'll',
'find', 'in', 'so', 'many', 'KL']

Filtered Sentence: ['Table', '&', 'Apron', '–', 'formerly', 'The', 'Kitchen',
'Table', 'Restaurant', '&', 'Bakery', '–', '’', 'exist', 'disrupt', 'scene',
'.', 'From', 'outside', ',', 'barely', 'stretches', 'boundaries', 'already',
'saturated', 'restaurant-cum-bakery', 'scene', '.', 'But', 'none', 'matters',
'.', 'Because', 'right', 'birth', '2014', ',', 'Table', '&', 'Apron', 'prove
n', 'restaurant', 'spades', 'component', 'elementary', 'yet', 'rare', '–', 'h
eart', '.', 'Through', 'hard', 'work', ',', 'dedication', 'boring', 'old-fash
ioned', 'virtues', 'honest', 'operation', ',', 'owner', 'Marcus', 'Low', 'tea
m', 'carved', 'us', 'little', 'treasure', 'Damansara', 'Kim', '.', '(', 'Cred
it', 'must', 'also', 'given', 'former', 'co-owner', 'Mei', 'Wan', 'Tan', '.',
')', 'The', 'narcissism', '’', 'find', 'many', 'KL']
```

Lexicon Normalization

Lexicon normalization considers another type of noise in the text. For example, connection, connected, connecting word reduce to a common word "connect". It reduces derivationally related forms of a word to a common root word.

Lemmatization

Lemmatization reduces words to their base word, which is linguistically correct lemmas. It transforms root word with the use of vocabulary and morphological analysis. Lemmatization is usually more sophisticated than stemming. Stemmer works on an individual word without knowledge of the context. For example, The word "better" has "good" as its lemma. This thing will miss by stemming because it requires a dictionary look-up.

In [18]:
```
#Lexicon Normalization
#performing stemming and Lemmatization
from nltk.stem.wordnet import WordNetLemmatizer

lem = WordNetLemmatizer()

from nltk.stem.porter import PorterStemmer
stem = PorterStemmer()

word = "flying"
print("Lemmatized Word:",lem.lemmatize(word,"v"))
print("Stemmed Word:",stem.stem(word))
```

```
Lemmatized Word: fly
Stemmed Word: fli
```

Stemming

Stemming is a process of linguistic normalization, which reduces words to their word root word or chops off the derivational affixes. For example, connection, connected, connecting word reduce to a common word "connect".

In [19]:
```
# Stemming
from nltk.stem import PorterStemmer
from nltk.tokenize import sent_tokenize, word_tokenize

ps = PorterStemmer()

stemmed_words=[]
for w in filtered_sent:
    stemmed_words.append(ps.stem(w))

print("Filtered Sentence:",filtered_sent)
print("Stemmed Sentence:",stemmed_words)
```

```
Filtered Sentence: ['Table', '&', 'Apron', '–', 'formerly', 'The', 'Kitchen',
'Table', 'Restaurant', '&', 'Bakery', '–', '’', 'exist', 'disrupt', 'scene',
'.', 'From', 'outside', ',', 'barely', 'stretches', 'boundaries', 'already',
'saturated', 'restaurant-cum-bakery', 'scene', '.', 'But', 'none', 'matters',
'.', 'Because', 'right', 'birth', '2014', ',', 'Table', '&', 'Apron', 'prove
n', 'restaurant', 'spades', 'component', 'elementary', 'yet', 'rare', '–', 'h
eart', '.', 'Through', 'hard', 'work', ',', 'dedication', 'boring', 'old-fash
ioned', 'virtues', 'honest', 'operation', ',', 'owner', 'Marcus', 'Low', 'tea
m', 'carved', 'us', 'little', 'treasure', 'Damansara', 'Kim', '.', '(', 'Cred
it', 'must', 'also', 'given', 'former', 'co-owner', 'Mei', 'Wan', 'Tan', '.',
')', 'The', 'narcissism', '’', 'find', 'many', 'KL']
Stemmed Sentence: ['tabl', '&', 'apron', '–', 'formerli', 'the', 'kitchen',
'tabl', 'restaur', '&', 'bakeri', '–', '’', 'exist', 'disrupt', 'scene', '.',
'from', 'outsid', ',', 'bare', 'stretch', 'boundari', 'alreadi', 'satur', 're
staurant-cum-bakeri', 'scene', '.', 'but', 'none', 'matter', '.', 'becaus',
'right', 'birth', '2014', ',', 'tabl', '&', 'apron', 'proven', 'restaur', 'sp
ade', 'compon', 'elementari', 'yet', 'rare', '–', 'heart', '.', 'through', 'h
ard', 'work', ',', 'dedic', 'bore', 'old-fashion', 'virtu', 'honest', 'oper',
',', 'owner', 'marcu', 'low', 'team', 'carv', 'us', 'littl', 'treasur', 'dama
nsara', 'kim', '.', '(', 'credit', 'must', 'also', 'given', 'former', 'co-ow
n', 'mei', 'wan', 'tan', '.', ')', 'the', 'narciss', '’', 'find', 'mani', 'k
l']
```

POS Tagging

The primary target of Part-of-Speech(POS) tagging is to identify the grammatical group of a given word. Whether it is a NOUN, PRONOUN, ADJECTIVE, VERB, ADVERBS, etc. based on the context. POS Tagging looks for relationships within the sentence and assigns a corresponding tag to the word.

Reference: https://pythonspot.com/nltk-speech-tagging/ (https://pythonspot.com/nltk-speech-tagging/)

In [23]:
```
from nltk import pos_tag
# https://cs.nyu.edu/grishman/jet/guide/PennPOS.html

# print(nltk.pos_tag(words))

print(nltk.pos_tag(words))
```

```
[('Table', 'NNP'), ('&', 'CC'), ('Apron', 'NNP'), ('–', 'NNP'), ('formerly',
'RB'), ('The', 'DT'), ('Kitchen', 'NNP'), ('Table', 'NNP'), ('Restaurant', 'N
NP'), ('&', 'CC'), ('Bakery', 'NNP'), ('–', 'NNP'), ('doesn', 'NN'), ('’', 'N
NP'), ('t', 'VBZ'), ('exist', 'VBP'), ('to', 'TO'), ('disrupt', 'VB'), ('th
e', 'DT'), ('scene', 'NN'), ('.', '.'), ('From', 'IN'), ('the', 'DT'), ('outs
ide', 'NN'), (',', ','), ('it', 'PRP'), ('barely', 'RB'), ('stretches', 'VB
Z'), ('the', 'DT'), ('boundaries', 'NNS'), ('of', 'IN'), ('what', 'WP'), ('i
s', 'VBZ'), ('an', 'DT'), ('already', 'RB'), ('saturated', 'VBN'), ('restaura
nt-cum-bakery', 'JJ'), ('scene', 'NN'), ('.', '.'), ('But', 'CC'), ('none',
'NN'), ('of', 'IN'), ('it', 'PRP'), ('matters', 'NNS'), ('.', '.'), ('Becaus
e', 'IN'), ('right', 'NN'), ('from', 'IN'), ('its', 'PRP$'), ('birth', 'NN'),
('in', 'IN'), ('2014', 'CD'), (',', ','), ('Table', 'NNP'), ('&', 'CC'), ('Ap
ron', 'NNP'), ('has', 'VBZ'), ('proven', 'VBN'), ('to', 'TO'), ('be', 'VB'),
('a', 'DT'), ('restaurant', 'NN'), ('that', 'WDT'), ('has', 'VBZ'), ('in', 'I
N'), ('spades', 'NNS'), ('a', 'DT'), ('component', 'NN'), ('so', 'RB'), ('ele
mentary', 'JJ'), ('yet', 'RB'), ('so', 'RB'), ('rare', 'JJ'), ('–', 'JJ'),
('heart', 'NN'), ('.', '.'), ('Through', 'IN'), ('hard', 'JJ'), ('work', 'N
N'), (',', ','), ('dedication', 'NN'), ('and', 'CC'), ('all', 'PDT'), ('the',
'DT'), ('boring', 'JJ'), ('old-fashioned', 'JJ'), ('virtues', 'NNS'), ('of',
'IN'), ('an', 'DT'), ('honest', 'NN'), ('operation', 'NN'), (',', ','), ('own
er', 'NN'), ('Marcus', 'NNP'), ('Low', 'NNP'), ('and', 'CC'), ('his', 'PRP
$'), ('team', 'NN'), ('have', 'VBP'), ('carved', 'VBN'), ('for', 'IN'), ('u
s', 'PRP'), ('a', 'DT'), ('little', 'JJ'), ('treasure', 'NN'), ('in', 'IN'),
('Damansara', 'NNP'), ('Kim', 'NNP'), ('.', '.'), ('(', '('), ('Credit', 'NN
P'), ('must', 'MD'), ('also', 'RB'), ('be', 'VB'), ('given', 'VBN'), ('to',
'TO'), ('former', 'JJ'), ('co-owner', 'NN'), ('Mei', 'NNP'), ('Wan', 'NNP'),
('Tan', 'NNP'), ('.', '.'), (')', ')'), ('The', 'DT'), ('narcissism', 'NN'),
('you', 'PRP'), ('’', 'VBP'), ('ll', 'JJ'), ('find', 'VBP'), ('in', 'IN'),
('so', 'RB'), ('many', 'JJ'), ('KL', 'NNP')]
```

In [24]:
```
text = """ Dostoevsky was the son of a doctor.
His parents were very hard-working and deeply religious people,
but so poor that they lived with their five children in only
two rooms. The father and mother spent their evenings
in reading aloud to their children, generally from books of
a serious character."""

words = word_tokenize(text)

import nltk
tagged = nltk.pos_tag(words)
print(tagged)
```

```
[('Dostoevsky', 'NNP'), ('was', 'VBD'), ('the', 'DT'), ('son', 'NN'), ('of',
'IN'), ('a', 'DT'), ('doctor', 'NN'), ('.', '.'), ('His', 'PRP$'), ('parent
s', 'NNS'), ('were', 'VBD'), ('very', 'RB'), ('hard-working', 'JJ'), ('and',
'CC'), ('deeply', 'RB'), ('religious', 'JJ'), ('people', 'NNS'), (',', ','),
('but', 'CC'), ('so', 'RB'), ('poor', 'JJ'), ('that', 'IN'), ('they', 'PRP'),
('lived', 'VBD'), ('with', 'IN'), ('their', 'PRP$'), ('five', 'CD'), ('childr
en', 'NNS'), ('in', 'IN'), ('only', 'RB'), ('two', 'CD'), ('rooms', 'NNS'),
('.', '.'), ('The', 'DT'), ('father', 'NN'), ('and', 'CC'), ('mother', 'NN'),
('spent', 'VBN'), ('their', 'PRP$'), ('evenings', 'NNS'), ('in', 'IN'), ('rea
ding', 'VBG'), ('aloud', 'NN'), ('to', 'TO'), ('their', 'PRP$'), ('children',
'NNS'), (',', ','), ('generally', 'RB'), ('from', 'IN'), ('books', 'NNS'),
('of', 'IN'), ('a', 'DT'), ('serious', 'JJ'), ('character', 'NN'), ('.',
'.')]
```

In [25]:
```
import re

stop_words = set(stopwords.words("english"))

text = """ Dostoevsky was the son of a doctor.
His parents were very hard-working and deeply religious people,
but so poor that they lived with their five children in only
two rooms. The father and mother spent their evenings
in reading aloud to their children, generally from books of
a serious character."""

words = word_tokenize(text)

def is_ok(token):
    return re.match('^[a-z]+$', token) and token not in stop_words

filtered = [word for word in word_tokenize(text.lower()) if is_ok(word)]

import nltk
tagged = nltk.pos_tag(filtered)
print(tagged)
```

```
[('dostoevsky', 'JJ'), ('son', 'NN'), ('doctor', 'NN'), ('parents', 'NNS'),
('deeply', 'RB'), ('religious', 'JJ'), ('people', 'NNS'), ('poor', 'JJ'), ('l
ived', 'VBD'), ('five', 'CD'), ('children', 'NNS'), ('two', 'CD'), ('rooms',
'NNS'), ('father', 'RB'), ('mother', 'RB'), ('spent', 'JJ'), ('evenings', 'NN
S'), ('reading', 'VBG'), ('aloud', 'JJ'), ('children', 'NNS'), ('generally',
'RB'), ('books', 'NNS'), ('serious', 'JJ'), ('character', 'NN')]
```

In [44]:
```
from nltk.corpus import brown
brown_tagged_sents = brown.tagged_sents(categories='news')
brown_sents = brown.sents(categories='news')

print(brown_tagged_sents)
```

```
[[('The', 'AT'), ('Fulton', 'NP-TL'), ('County', 'NN-TL'), ('Grand', 'JJ-T
L'), ('Jury', 'NN-TL'), ('said', 'VBD'), ('Friday', 'NR'), ('an', 'AT'), ('in
vestigation', 'NN'), ('of', 'IN'), ("Atlanta's", 'NP$'), ('recent', 'JJ'),
('primary', 'NN'), ('election', 'NN'), ('produced', 'VBD'), ('``', '``'), ('n
o', 'AT'), ('evidence', 'NN'), ("''", "''"), ('that', 'CS'), ('any', 'DTI'),
('irregularities', 'NNS'), ('took', 'VBD'), ('place', 'NN'), ('.', '.')],
[('The', 'AT'), ('jury', 'NN'), ('further', 'RBR'), ('said', 'VBD'), ('in',
'IN'), ('term-end', 'NN'), ('presentments', 'NNS'), ('that', 'CS'), ('the',
'AT'), ('City', 'NN-TL'), ('Executive', 'JJ-TL'), ('Committee', 'NN-TL'),
(',', ','), ('which', 'WDT'), ('had', 'HVD'), ('over-all', 'JJ'), ('charge',
'NN'), ('of', 'IN'), ('the', 'AT'), ('election', 'NN'), (',', ','), ('``', ''
``'), ('deserves', 'VBZ'), ('the', 'AT'), ('praise', 'NN'), ('and', 'CC'), ('t
hanks', 'NNS'), ('of', 'IN'), ('the', 'AT'), ('City', 'NN-TL'), ('of', 'IN-T
L'), ('Atlanta', 'NP-TL'), ("''", "''"), ('for', 'IN'), ('the', 'AT'), ('mann
er', 'NN'), ('in', 'IN'), ('which', 'WDT'), ('the', 'AT'), ('election', 'N
N'), ('was', 'BEDZ'), ('conducted', 'VBN'), ('.', '.')], ...]
```

In [30]:
```
import nltk
from nltk.tokenize import word_tokenize
from nltk import chunk

text = "The white dog fight with a black cat"
words = word_tokenize(text)
tagged = nltk.pos_tag(words)

grammar = "NP: {<DT>?<JJ>*<NN>}"
cp = nltk.RegexpParser(grammar)
result = cp.parse(tagged)
print(result)
result.draw()
```

```
(S
  (NP The/DT white/JJ dog/NN)
  (NP fight/NN)
  with/IN
  (NP a/DT black/JJ cat/NN))
```

In [35]:
```
from nltk import ne_chunk

def entities(text):
    return ne_chunk(
        pos_tag(
            word_tokenize(text)
        )
    )

tree = entities (text)

tree.pprint()
```

```
(S The/DT white/JJ dog/NN fight/NN with/IN a/DT black/JJ cat/NN)
```

In [36]:
```
#tree.draw

#from nltk.corpus import treebank
#tree = treebank.parsed_sents('wsj_0001.mrg')[0]
tree.draw
```

Out[36]:
```
<bound method Tree.draw of Tree('S', [('The', 'DT'), ('white', 'JJ'), ('dog',
'NN'), ('fight', 'NN'), ('with', 'IN'), ('a', 'DT'), ('black', 'JJ'), ('cat',
'NN')])>
```

In [39]:
```
# Step 1: Load Data
from nltk.corpus import names
labeled_names = ([(name, 'male') for name in names.words('male.txt')]
    + [(name, 'female') for name in names.words('female.txt')])

import random
random.shuffle(labeled_names)
#print(labeled_names[:10])

# Step 2: Extract last letter of a name as the feature and form feature set
def feature_extractor(name):
    return {'last_letter': name[-1]}

featureset = [(feature_extractor(name), gender) for (name, gender) in labeled_r
# print(featureset[:10])

# Step 3: Split the feature set to training/testing datasets

train_set, test_set = featureset[500:], featureset[:500]

# Step 4/5: Load the classifier and perform training
import nltk
classifier = nltk.NaiveBayesClassifier.train(train_set)

# Step 6: Prediction/Evaluation
print(classifier.classify(feature_extractor('Danny')))
print(nltk.classify.accuracy(classifier, test_set))
```

```
female
0.762
```

In [42]:
```
import nltk

# Step 1: Load Data
from nltk.corpus import movie_reviews
documents = [(list(movie_reviews.words(fileid)), category)
             for category in movie_reviews.categories()
             for fileid in movie_reviews.fileids(category)]

import random
random.shuffle(documents)

# Step 2: Extract Feature
all_words = []
for w in movie_reviews.words():
    all_words.append(w.lower())
all_words = nltk.FreqDist(all_words)
word_features = list(all_words.keys())[:3000]

def feature_extractor(review):
    words = set(review)
    features = {}
    for w in word_features:
        features[w] = (w in words)

    return features

featureset = [(feature_extractor(review), sentiment) for (review, sentiment) ir


# Step 3: Split the feature set to training/testing datasets
training_set, testing_set = featureset[:1900],featureset[1900:]

# Step 4/5: Load the classifier and perform training
classifier = nltk.NaiveBayesClassifier.train(training_set)

# Step 6: Prediction/Evaluation
print("Classifier accuracy:",nltk.classify.accuracy(classifier, testing_set))
classifier.show_most_informative_features(15)
```

```
Classifier accuracy: 0.85
Most Informative Features
                   sucks = True              neg : pos    =     10.1 : 1.0
                  annual = True              pos : neg    =      9.7 : 1.0
                 frances = True              pos : neg    =      8.4 : 1.0
            unimaginative = True              neg : pos    =      8.3 : 1.0
                  turkey = True              neg : pos    =      7.8 : 1.0
             silverstone = True              neg : pos    =      7.6 : 1.0
              schumacher = True              neg : pos    =      7.4 : 1.0
                 idiotic = True              neg : pos    =      7.2 : 1.0
                  regard = True              pos : neg    =      7.0 : 1.0
                  alicia = True              neg : pos    =      7.0 : 1.0
               atrocious = True              neg : pos    =      7.0 : 1.0
                    mena = True              neg : pos    =      7.0 : 1.0
                  shoddy = True              neg : pos    =      7.0 : 1.0
                  suvari = True              neg : pos    =      7.0 : 1.0
                 cunning = True              pos : neg    =      6.4 : 1.0
```

TextBlob

Installation

conda install -c conda-forge textblob

Difference between TextBlob and NLTK

Reference: https://www.quora.com/What-is-the-use-of-NLTK-and-TextBlob-What-is-the-difference-between-both-And-for-text-analysis-which-tool-is-better#:~:text=I%20als-,NLTK%20and%20TextBlob%20are%20both%20excellent%20libraries%2((https://www.quora.com/What-is-the-use-of-NLTK-and-TextBlob-What-is-the-difference-between-both-And-for-text-analysis-which-tool-is-better#:~:text=I%20als-,NLTK%20and%20TextBlob%20are%20both%20excellent%20libraries%2(

In [43]:
```
from textblob import TextBlob

blob = TextBlob(text)

print(blob.noun_phrases)
```

```
---------------------------------------------------------------------------
ModuleNotFoundError                       Traceback (most recent call last)
Cell In[43], line 1
----> 1 from textblob import TextBlob
      3 blob = TextBlob(text)
      5 print(blob.noun_phrases)

ModuleNotFoundError: No module named 'textblob'
```

Reference: https://www.datacamp.com/community/tutorials/text-analytics-beginners-nltk (https://www.datacamp.com/community/tutorials/text-analytics-beginners-nltk)

Quiz

Exploring Features of NLTK:

a. Open the text file for processing: First, we are going to open and read the file which we want to analyze eg. The fishing documentation in txt file (page2). Reference: https://huntfish.mdc.mo.gov/sites/default/files/downloads/page/IntroToFishing_2017_v2.pdf (https://huntfish.mdc.mo.gov/sites/default/files/downloads/page/IntroToFishing_2017_v2.pdf)

b. Import required libraries: For various data processing cases in NLP, we need to import some libraries. In this case, we are going to use NLTK for Natural Language Processing. We will use it to perform various operations on the text.

c. Sentence tokenizing: By tokenizing the text with sent_tokenize( ), we can get the text as sentences.

d. Word tokenizing: By tokenizing the text with word_tokenize( ), we can get the text as words.

e. Find the frequency distribution: Let's find out the frequency of words in our text.

f. Plot the frequency graph: Let's plot a graph to visualize the word distribution in our text.

g. Remove punctuation marks: Next, we are going to remove the punctuation marks as they are not very useful for us. We are going to use isalpha( ) method to separate the punctuation marks from the actual text. Also, we are going to make a new list called words_no_punc, which will store the words in lower case but exclude the punctuation marks.

h. Plotting graph without punctuation marks:

i. List of stopwords:

j. Removing stopwords:

k. Final frequency distribution: the final graph has many useful words that help us understand what our sample data is about, showing how essential it is to perform data cleaning on NLP.

Hint: https://medium.com/towards-artificial-intelligence/natural-language-processing-nlp-with-python-tutorial-for-beginners-1f54e610a1a0 (https://medium.com/towards-artificial-intelligence/natural-language-processing-nlp-with-python-tutorial-for-beginners-1f54e610a1a0)

In [62]:
```
# b

import nltk

nltk.download('punkt')
nltk.download('stopwords')
nltk.download("wordnet")
nltk.download('averaged_perceptron_tagger')
nltk.download('brown')
nltk.download('maxent_ne_chunker')
nltk.download('words')
nltk.download('names')
nltk.download('movie_reviews')
```

```
[nltk_data] Downloading package punkt to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package punkt is already up-to-date!
[nltk_data] Downloading package stopwords to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
[nltk_data] Downloading package wordnet to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package wordnet is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package averaged_perceptron_tagger is already up-to-
[nltk_data]       date!
[nltk_data] Downloading package brown to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package brown is already up-to-date!
[nltk_data] Downloading package maxent_ne_chunker to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package maxent_ne_chunker is already up-to-date!
[nltk_data] Downloading package words to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package words is already up-to-date!
[nltk_data] Downloading package names to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package names is already up-to-date!
[nltk_data] Downloading package movie_reviews to
[nltk_data]     C:\Users\Asus\AppData\Roaming\nltk_data...
[nltk_data]   Package movie_reviews is already up-to-date!
```

Out[62]: True

In [66]:
```
# a
text = open('./An_Introduction_to_Fishing.txt', encoding='utf-8').read()
text = text.replace ('  ','')
text = text.replace('\n', '')
text = text.replace('\x0c', ' ')
text
```

Out[66]:
```
" An Introduction toFISHING Fishing “God never did make a more calm, F ishi
ng is a great way to spenda day. You can take a lunch andpicnic as you fsh.
You can camp quiet, innocent recreation ....”Izaak Waltonnear a lake. You c
an hike or boat toa fshing spot. Many people buildtheir vacations around fs
hing. Besure to take your family or yourfriends along, for there is no more
sociable activity.Missouri has more than 800,000acres of surface water, and
mostof it provides great fshing. Ourwaters hold ancient paddlefsh,wary larg
emouth bass, and tastybluegill—more than 200 diferentspecies. About 40 of t
hose fshspecies are the targets of anglers.Some Missourians fsh for sportor
relaxation, while others fsh onlyfor food. Regardless of motivation,the maj
ority of anglers reap all thebenefts of fshing. They spendquality time on t
he water and thenreturn home to a satisfying mealof fried or grilled fsh th
ey havecaught themselves.One of the joys of fshing is thatit can be fun and
productive atany skill level. You can complicatethe sport with jargon andso
phisticated equipment, but thewhole sport can be pared downto some basic eq
uipment andtechniques.This publication presents thosebasics to you. It woul
d take manyvolumes to introduce you to all thespecies of fsh, all the metho
ds,and all the lures, and so muchinformation at once would beoverwhelming.I
nstead, we want to be yourguide for your frst fshing trips. Letus go with y
```

In [67]:
```
# c
sentences = nltk.sent_tokenize(text)
sentences
```

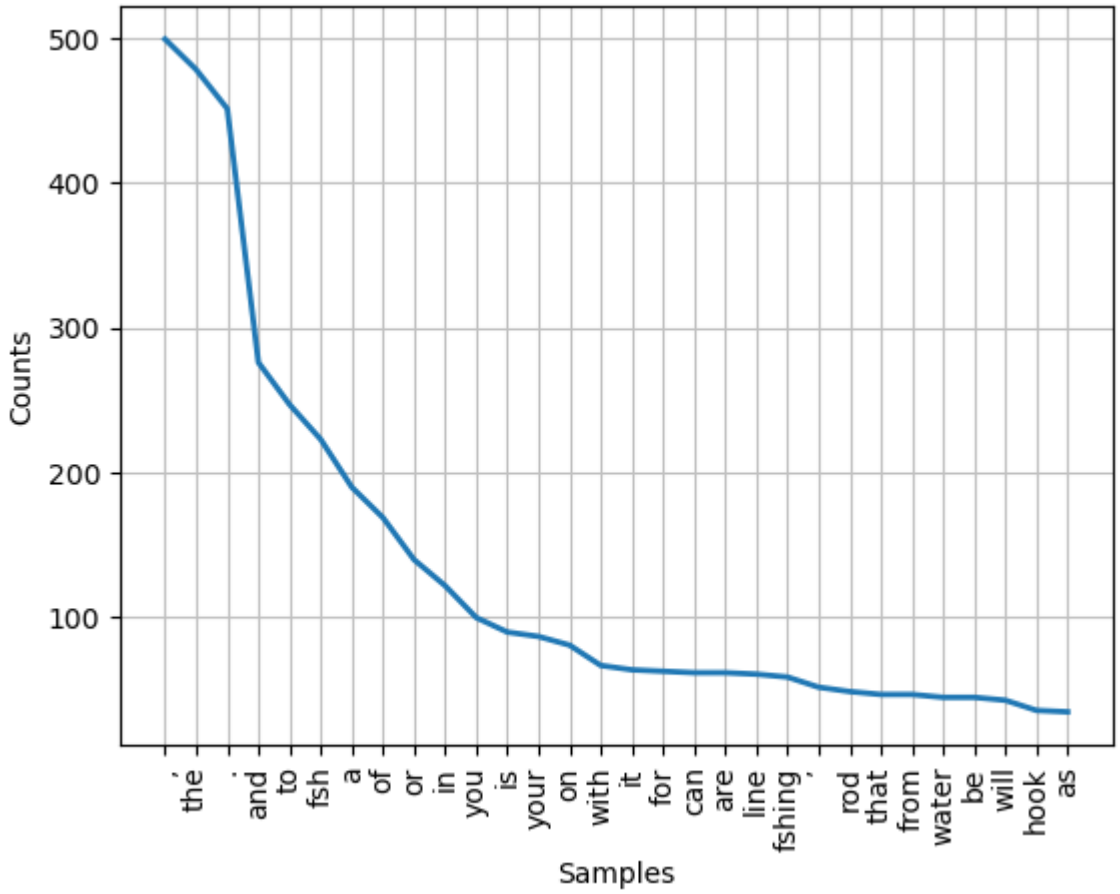
Out[67]:
```
[' An Introduction toFISHING Fishing “God never did make a more calm, F ish
ing is a great way to spenda day.',
 'You can take a lunch andpicnic as you fsh.',
 'You can camp quiet, innocent recreation ....”Izaak Waltonnear a lake.',
 'You can hike or boat toa fshing spot.',
 'Many people buildtheir vacations around fshing.',
 'Besure to take your family or yourfriends along, for there is no moresoci
able activity.Missouri has more than 800,000acres of surface water, and mos
tof it provides great fshing.',
 'Ourwaters hold ancient paddlefsh,wary largemouth bass, and tastybluegill—
more than 200 diferentspecies.',
 'About 40 of those fshspecies are the targets of anglers.Some Missourians
fsh for sportor relaxation, while others fsh onlyfor food.',
 'Regardless of motivation,the majority of anglers reap all thebenefts of f
shing.',
 'They spendquality time on the water and thenreturn home to a satisfying m
ealof fried or grilled fsh they havecaught themselves.One of the joys of fs
hing is thatit can be fun and productive atany skill level.',
 'You can complicatethe sport with jargon andsophisticated equipment, but t
```

In [68]:
```
# d
words = word_tokenize(text)
words
```

Out[68]:
```
['An',
 'Introduction',
 'toFISHING',
 'Fishing',
 '“',
 'God',
 'never',
 'did',
 'make',
 'a',
 'more',
 'calm',
 ',',
 'F',
 'ishing',
 'is',
 'a',
 'great',
 'way',
```

In [84]:
```
# e

import matplotlib.pyplot as plt
import string
stop_words.update(set(string.punctuation))
fdist= FreqDist(words)
print(fdist)

# f
fdist.plot(30,cumulative=False)

plt.show()

# g h i j
tokenized_sent = words

filtered_sent=[]
for w in tokenized_sent:
    if w not in stop_words and w!='’':
        filtered_sent.append(w)

fdist2= FreqDist(filtered_sent)
# k
fdist2.plot(30,cumulative=False)

plt.show()
```

```
<FreqDist with 2818 samples and 10035 outcomes>
```



```
[('fsh', 223), ('line', 61), ('fshing', 59), ('rod', 49), ('water', 45)]
```