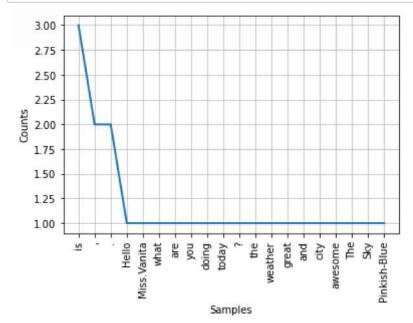
```
In [1]: #Loading NLTK
         #TEXT MINING ANALYSIS
         #1.NLTK IS A POWERFUL PACKAGE THAT PROVIDES A SET OF DIVERSE NATURAL LANGUAGES AL
         #2.IT IS FREE, OPENSOURCE EASY TO USE AND WEEL DOCUMENTED.
         #3.NLTK CONSISTS OF THE MOST COMMON ALGORITHMS SUCH AS TOKENZING, PART OD SPEECH 1
         # TOPIC SEGMENTATION, AND NAMED ENTITY RECOGNITION NLTK HELPS THE COMPUTER TO ANAL
         import nltk
In [10]: #Tokenization is the first step in Text Analytics.
         #The Process of Breaking Down a Text Paragraph into Smaller Chunks Such as Words
         #Token is Single Entity That is Building Blocks For Sentence or Paragraph.
         #SENTENCE TOKENIZATION
         from nltk.tokenize import sent_tokenize
         text="""Hello Miss. Vanita, what are you doing today? the weather is great, and city
         tokenized_sent=sent_tokenize(text)
         print(tokenized sent)
         ['Hello Miss.Vanita, what are you doing today?', 'the weather is great, and city
          is awesome.', 'The Sky is Pinkish-Blue.']
In [11]: # Word Tokenizer Breaks Text Paragraph into Words.
         # WORD TOKENIZATION
         from nltk.tokenize import word tokenize
         text="""Hello Miss. Vanita, what are you doing today? the weather is great, and city
         tokenized word=word tokenize(text)
         print(tokenized word)
          ['Hello', 'Miss.Vanita', ',', 'what', 'are', 'you', 'doing', 'today', '?', 'th
         e', 'weather', 'is', 'great', ',', 'and', 'city', 'is', 'awesome', '.', 'The', 'Sky', 'is', 'Pinkish-Blue', '.']
 In [8]: #FREQUENCY DISTRIBUTION
         from nltk.probability import FreqDist
         fdist=FreqDist(tokenized_word)
         print(fdist)
          <FreqDist with 20 samples and 24 outcomes>
 In [6]: | fdist.most common(2)
 Out[6]: [('is', 3), (',', 2)]
```

## In [9]: #FREQUENCY DISTRIBUTION PLOT import matplotlib.pyplot as plt fdist.plot(30,cumulative=False) plt.show()



```
In [10]: import nltk
    nltk.download('punkt')
    nltk.download('wordnet')

        [nltk_data] Downloading package punkt to
        [nltk_data] C:\Users\Owner\AppData\Roaming\nltk_data...
        [nltk_data] Package punkt is already up-to-date!
        [nltk_data] Downloading package wordnet to
        [nltk_data] C:\Users\Owner\AppData\Roaming\nltk_data...

Out[10]: True

In [22]: nltk.word_tokenize("hi How are you")

Out[22]: ['hi', 'How', 'are', 'you']
```

#### In [12]: #STOPWORDS CONSIDERED AS NOISE IN THE TEXT.TEXT MAY CONTAIN STOP WORDS SUCH AS IS #STOPWORDS

from nltk.corpus import stopwords
stop\_words=stopwords.words("english")
print(stop words)

['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", "you'll", "you'd", 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "i t's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'i s', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'havin g', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about', 'a gainst', 'between', 'into', 'through', 'during', 'before', 'after', 'above', 'b elow', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'h ow', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'suc h', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "could n't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "must n't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wa sn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't"]

### In [23]: print(len(stopwords)) print(stopwords)

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{'now', 'as', 'weren', 'of', 'between', "aren't", "didn't", 'once', 'won', 'abo ut', "that'll", 'yourself', 'i', 'some', 'needn', 'through', 'and', 'again', 'a in', 'own', "wouldn't", 'were', 'further', 'mightn', 'these', 'both', 'll', 'yo ur', 'with', "you've", 'all', 'too', 'y', 'any', 'into', 'or', 'herself', 'at', 'down', 'shouldn', 'in', 'what', "hadn't", 'the', 'shan', 'such', 'during', 'o', 'nor', 'being', "you'd", 'was', 'above', 'who', 'after', 'there', 'for', 'did', "couldn't", 'm', 'it', 'just', 't', "don't", 'which', 'him', 'doesn', 'y ou', 'my', 'more', 'on', "shouldn't", 'so', 'will', 'no', 'this', 'by', 'we', 'having', "haven't", 'myself', 're', 'where', 's', 'didn', 'yourselves', 'had n', 'ma', 'to', 'its', 'himself', 'is', 'ours', "mustn't", 'other', 'if', 'fro m', 'ourselves', 'than', 'not', 'aren', "isn't", 'up', 'under', 'most', 'wasn', 'hasn', 'me', 'should', "should've", 'be', 'are', "hasn't", "mightn't", 'hers', 'because', 'been', 'have', 'wouldn', 'each', 'they', "weren't", "she's", 'thos e', "you'll", 'she', 'has', 'mustn', 'their', 'yours', 'our', 'itself', 'when', 'below', 'does', 'why', 'an', 'few', 'off', "wasn't", 'whom', "needn't", 'a', 'd', 'can', 'very', "doesn't", 'doing', 'don', 'her', 'themselves', 'isn', 'ha d', 'over', "shan't", 'while', 'against', 'them', 'am', 'he', 'how', 'same', 'h aven', 'until', 'before', 'then', 'only', "won't", 'that', "you're", "it's", 't heirs', 'here', 've', 'do', 'couldn', 'out', 'but', 'his'}

```
In [18]: stop_words.append('work')
print(stop_words)
```

['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", "you'll", "you'd", 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "i t's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'i s', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'havin g', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about', 'a gainst', 'between', 'into', 'through', 'during', 'before', 'after', 'above', 'b elow', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'h ow', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'suc h', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "could n't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "wustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't", 'work']

# In [21]: # Removing Stopwords from nltk.tokenize import sent\_tokenize,word\_tokenize from nltk.corpus import stopwords data="AI was introduced in the year 1956 but it gained popularity recently." stopwords=set(stopwords.words('english')) words=word\_tokenize(data) wordsFiltered=[] for w in words: if w not in stopwords: wordsFiltered.append(w) print(wordsFiltered)

['AI', 'introduced', 'year', '1956', 'gained', 'popularity', 'recently', '.']

```
In [15]: #Stemming is The Process of Bringing Words Back to Their Root form This Way You &
          #For Example: Connection, Connected, Connected Word Reduce to a Common Word 'Conne
          import nltk
          from nltk.stem import PorterStemmer
          #from nltk.tokenize import word_tokenize
          stemmer= PorterStemmer()
          Input_str="There are several types of stemming Algorithms."
          Input_str=nltk.word_tokenize(Input_str)
          for word in Input_str:
              print(stemmer.stem(word))
          there
          are
          sever
          type
          of
          stem
          algorithm
 In [6]: #Lemmatization is Same Like Stemming i.e it Hve Same Goals As Like Stemming But L
          import nltk
          wn=nltk.WordNetLemmatizer()
          ps=nltk.PorterStemmer()
         dir(wn)
Out[6]: ['__class__',
              _delattr___',
             _dict__'
             _
_dir__',
             _doc__
              _eq__',
              _format___',
             _ge__',
             _getattribute__',
             _gt__',
             _hash___',
             _init___',
              _init_subclass___',
             _le__',
            __lt___'
             _module___',
             _ne__',
             _new__',
             _reduce__',
             _reduce_ex__',
             _repr__',
             _setattr__',
             _sizeof__',
             _str__',
              _subclasshook___',
            __weakref__',
           'lemmatize']
```

```
In [38]: print(ps.stem('goose'))
         print(ps.stem('geese'))
         goos
         gees
In [39]:
         print(wn.lemmatize('cactus'))
         print(wn.lemmatize('cacti'))
         cactus
         cactus
In [16]: #Stemming Code
         import nltk
         from nltk.stem.porter import PorterStemmer
         porter_stemmer=PorterStemmer()
         text="studies studying cries cry"
         tokenization=nltk.word tokenize(text)
         for w in tokenization:
              print("Stemming for {} is {}".format(w,porter_stemmer.stem(w)))
         Stemming for studies is studi
         Stemming for studying is studi
         Stemming for cries is cri
         Stemming for cry is cri
In [17]: # Lemmatization Code
         import nltk
         from nltk.stem import WordNetLemmatizer
         wordnet lemmatizer=WordNetLemmatizer()
         text="studies studying cries cry"
         tokenization=nltk.word tokenize(text)
         for w in tokenization:
              print("lemma for {} is {}".format(w,wordnet_lemmatizer.lemmatize(w)))
         lemma for studies is study
         lemma for studying is studying
         lemma for cries is cry
         lemma for cry is cry
In [24]: # POS Tagging(Part of Speech Tagging) is The Process of attributing a Grammatical
         import nltk
         text=nltk.word tokenize("It is a pleasant day today")
         nltk.pos_tag(text)
Out[24]: [('It', 'PRP'),
          ('is', 'VBZ'),
('a', 'DT'),
          ('pleasant', 'JJ'),
           ('day', 'NN'),
           ('today', 'NN')]
```

```
In [34]: | nltk.help.upenn tagset('NNS')
         NNS: noun, common, plural
             undergraduates scotches bric-a-brac products bodyguards facets coasts
             divestitures storehouses designs clubs fragrances averages
             subjectivists apprehensions muses factory-jobs ...
In [38]: | nltk.help.upenn_tagset('VB,*')
         VB: verb, base form
             ask assemble assess assign assume atone attention avoid bake balkanize
             bank begin behold believe bend benefit bevel beware bless boil bomb
             boost brace break bring broil brush build ...
         VBD: verb, past tense
             dipped pleaded swiped regummed soaked tidied convened halted registered
             cushioned exacted snubbed strode aimed adopted belied figgered
             speculated wore appreciated contemplated ...
         VBG: verb, present participle or gerund
             telegraphing stirring focusing angering judging stalling lactating
             hankerin' alleging veering capping approaching traveling besieging
             encrypting interrupting erasing wincing ...
         VBN: verb, past participle
             multihulled dilapidated aerosolized chaired languished panelized used
             experimented flourished imitated reunifed factored condensed sheared
             unsettled primed dubbed desired ...
         VBP: verb, present tense, not 3rd person singular
             predominate wrap resort sue twist spill cure lengthen brush terminate
             appear tend stray glisten obtain comprise detest tease attract
             emphasize mold postpone sever return wag ...
         VBZ: verb, present tense, 3rd person singular
             bases reconstructs marks mixes displeases seals carps weaves snatches
             slumps stretches authorizes smolders pictures emerges stockpiles
             seduces fizzes uses bolsters slaps speaks pleads ...
         import nltk
In [39]:
         text=nltk.word_tokenize("I cannot bear the pain of bear")
         nltk.pos tag(text)
Out[39]: [('I', 'PRP'),
          ('can', 'MD'),
          ('not', 'RB'),
          ('bear', 'VB'),
          ('the', 'DT'),
          ('pain', 'NN'),
          ('of', 'IN'),
          ('bear', 'NN')]
In [51]: # Bag of Words: Bag of Words is The Simplest Way of Structuring Textual data Ever
         import sklearn
         from sklearn.feature_extraction.text import CountVectorizer
```

```
In [48]: phrases=["the quick brown fox jumped over the lazy dog"]
In [46]: | vect = CountVectorizer()
         vect.fit(phrases)
Out[46]: CountVectorizer()
In [47]: | print("Vocabulary size: {}".format(len(vect.vocabulary_)))
         print("Vocabulary content:\n {}".format(vect.vocabulary_))
         Vocabulary size: 8
         Vocabulary content:
          {'the': 7, 'quick': 6, 'brown': 0, 'fox': 2, 'jumped': 3, 'over': 5, 'lazy':
         4, 'dog': 1}
In [49]: bag_of_words = vect.transform(phrases)
In [50]: print(bag of words)
           (0, 0)
                         1
           (0, 1)
                          1
                         1
           (0, 2)
           (0, 3)
                         1
           (0, 4)
                         1
           (0, 5)
                         1
           (0, 6)
                         1
                          2
           (0, 7)
In [52]: print("bag_of_words as an array:\n{}".format(bag_of_words.toarray()))
         bag_of_words as an array:
         [[1 1 1 1 1 1 1 2]]
In [53]: vect.get feature names()
Out[53]: ['brown', 'dog', 'fox', 'jumped', 'lazy', 'over', 'quick', 'the']
In [ ]:
```