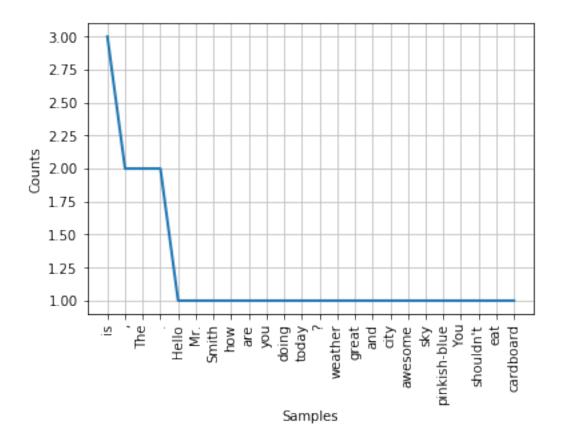
```
[]: import nltk
      nltk.download('punkt')
[21]: from nltk.tokenize import sent_tokenize
[22]: text="""Hello Mr. Smith, how are you doing today? The weather is
      great, and city is awesome. The sky is pinkish-blue. You shouldn't
      eat cardboard"""
      tokenized_text=sent_tokenize(text)
      print(tokenized_text)
     ['Hello Mr. Smith, how are you doing today?', 'The weather is\ngreat, and city
     is awesome.', 'The sky is pinkish-blue.', "You shouldn't\neat cardboard"]
[23]: from nltk.tokenize import word_tokenize
      tokenize_word=word_tokenize(text)
      print(tokenize_word)
     ['Hello', 'Mr.', 'Smith', ',', 'how', 'are', 'you', 'doing', 'today', '?',
     'The', 'weather', 'is', 'great', ',', 'and', 'city', 'is', 'awesome', '.',
     'The', 'sky', 'is', 'pinkish-blue', '.', 'You', "shouldn't", 'eat', 'cardboard']
[24]: from nltk.probability import FreqDist
      fdist=FreqDist(tokenize_word)
      print(fdist)
     <FreqDist with 24 samples and 29 outcomes>
[25]: fdist.most_common(2)
[25]: [('is', 3), (',', 2)]
[26]: import matplotlib.pyplot as plt
      fdist.plot(30,cumulative=False)
      plt.show()
```



```
[27]: sent = "Albert Einstein was born in Ulm, Germany in 1879."
      tokens=nltk.word_tokenize(sent)
      print(tokens)
     ['Albert', 'Einstein', 'was', 'born', 'in', 'Ulm', ',', 'Germany', 'in', '1879',
     '.']
 []: nltk.download('averaged_perceptron_tagger')
[29]: nltk.pos_tag(tokens)
[29]: [('Albert', 'NNP'),
       ('Einstein', 'NNP'),
       ('was', 'VBD'),
       ('born', 'VBN'),
       ('in', 'IN'),
       ('Ulm', 'NNP'),
       (',', ','),
       ('Germany', 'NNP'),
       ('in', 'IN'),
       ('1879', 'CD'),
```

```
('.', '.')]
```

```
[]: import nltk
      nltk.download('stopwords')
[34]: from nltk.corpus import stopwords
      stop_words=set(stopwords.words("english"))
      print(stop_words)
     {'aren', 'was', 'needn', 'why', 'weren', 'yours', 'so', 'both', "hadn't",
     'they', "weren't", 'whom', 'is', 'do', 'until', 'again', 'should', 'doesn',
     'wouldn', 'm', 'into', 'on', 'because', 'being', 've', 'up', 't', 'his',
     'herself', 'y', 'when', 'which', 'no', "didn't", 'myself', "shan't",
     "shouldn't", 'some', 'that', 'once', 'before', 'what', 'most', 'mustn', 'of',
     'ma', 'any', 'your', "doesn't", 'each', 'shouldn', "you'd", 'down', 'such',
     'didn', 'them', 'isn', 'from', 'its', 'while', 'than', 'won', 'there', 'or',
     'we', 'above', 'hasn', "she's", 'to', 'having', 'here', 'own', 'were',
     'themselves', 'himself', 'will', 'had', 'same', "it's", 'd', 'doing', 'too',
     "you'll", 're', 'are', 'very', 'haven', 'for', 's', "wasn't", 'all', "mustn't",
     "wouldn't", 'below', 'after', "couldn't", 'yourselves', 'shan', 'through', 'in',
     'other', 'between', "you've", 'these', 'by', "won't", 'have', 'hadn', 'further',
     'only', 'at', 'itself', 'couldn', 'those', 'then', 'this', 'she', 'during',
     'don', 'me', 'more', 'mightn', 'been', 'does', 'few', 'now', 'o', 'but',
     'yourself', 'theirs', 'the', 'their', 'be', 'if', 'not', 'where', "hasn't",
     'an', "aren't", "haven't", 'about', 'can', 'ain', 'under', 'it', 'hers', 'i',
     'wasn', 'nor', 'how', 'ourselves', 'our', 'him', 'my', 'her', 'against',
     "mightn't", 'a', 'off', 'ours', 'll', "should've", 'as', 'has', "you're", 'you',
     "don't", 'and', 'did', 'am', 'who', "that'll", 'out', 'with', "needn't", 'over',
     'just', "isn't", 'he'}
[38]: filtered sent=[]
      for w in tokenize_word:
          if w not in stop_words:
              filtered_sent.append(w)
      print("Tokenized Sentence:",tokenize_word)
      print('\n')
      print("Filterd Sentence:",filtered_sent)
     Tokenized Sentence: ['Hello', 'Mr.', 'Smith', ',', 'how', 'are', 'you', 'doing',
     'today', '?', 'The', 'weather', 'is', 'great', ',', 'and', 'city', 'is',
     'awesome', '.', 'The', 'sky', 'is', 'pinkish-blue', '.', 'You', "shouldn't",
     'eat', 'cardboard']
     Filterd Sentence: ['Hello', 'Mr.', 'Smith', ',', 'today', '?', 'The', 'weather',
     'great', ',', 'city', 'awesome', '.', 'The', 'sky', 'pinkish-blue', '.', 'You',
     'eat', 'cardboard']
```

```
[40]: 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('\n')
      print("Stemmed Sentence:",stemmed_words)
     Filtered Sentence: ['Hello', 'Mr.', 'Smith', ',', 'today', '?', 'The',
     'weather', 'great', ',', 'city', 'awesome', '.', 'The', 'sky', 'pinkish-blue',
     '.', 'You', 'eat', 'cardboard']
     Stemmed Sentence: ['hello', 'mr.', 'smith', ',', 'today', '?', 'the', 'weather',
     'great', ',', 'citi', 'awesom', '.', 'the', 'sky', 'pinkish-blu', '.', 'you',
     'eat', 'cardboard']
 []: import nltk
      nltk.download('wordnet')
[44]: 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('\n')
      print("Stemmed Word:",stem.stem(word))
     Lemmatized Word: fly
     Stemmed Word: fli
[45]: import pandas as pd
      import sklearn as sk
      import math
      first_sentence = "Data Science is the sexiest job of the 21st century"
      second sentence = "machine learning is the key for data science"
      first_sentence= first_sentence.split(" ")
      second sentence = second sentence.split(" ")
      total=set(first_sentence).union(set(second_sentence))
      print(total)
     {'21st', 'for', 'is', 'of', 'sexiest', 'data', 'learning', 'Data', 'key',
     'century', 'machine', 'job', 'Science', 'the', 'science'}
```

```
[48]: wordDictA = dict.fromkeys(total, 0)
      wordDictB = dict.fromkeys(total, 0)
      for word in first_sentence:
      wordDictA[word]+=1
      for word in second_sentence:
      wordDictB[word]+=1
      pd.DataFrame([wordDictA, wordDictB])
[48]:
        21st for
                   is of sexiest
                                   data learning Data key
                                                               century machine \
            1
                0
                    1
                        1
                                  1
                                       0
                                                 0
                                                       1
                                                            0
                                                                     1
                                                                              0
      1
           0
                 1
                    1
                        0
                                 0
                                       1
                                                 1
                                                       0
                                                            1
                                                                     0
                                                                              1
         job Science the science
      0
                    1
                         2
      1
          0
                    0
                        1
                                 1
[57]: def computeTF(wordDict, doc):
      tfDict = {}
       corpusCount = len(doc)
       for word, count in wordDict.items():
           tfDict[word] = count/float(corpusCount)
      return(tfDict)
                                                               #running our sentences_
       ⇔through the tf function:
      tfFirst = computeTF(wordDictA, first_sentence)
      tfSecond = computeTF(wordDictB, second sentence)
                                                                            ш
       →#Converting to dataframe for visualization
      tf=pd.DataFrame([tfFirst,tfSecond])
      print(tf)
                        is
        21st
                for
                             of
                               sexiest
                                           data learning Data
                                                                   key century \
       0.1 0.000 0.100 0.1
                                     0.1 0.000
                                                    0.000
                                                            0.1 0.000
                                                                            0.1
         0.0 0.125 0.125 0.0
                                     0.0 0.125
                                                    0.125
                                                            0.0 0.125
                                                                            0.0
        machine job Science
                                 the science
          0.000 0.1
                          0.1 0.200
                                        0.000
          0.125 0.0
                          0.0 0.125
                                        0.125
[58]: from nltk.corpus import stopwords
      stop_words =set(stopwords.words('english'))
      filtered_sentence = [w for w in wordDictA if not w in stop_words]
      print(filtered_sentence)
     ['21st', 'sexiest', 'data', 'learning', 'Data', 'key', 'century', 'machine',
     'job', 'Science', 'science']
```

```
[62]: def computeIDF(docList):
          idfDict = {}
          N = len(docList)
          idfDict = dict.fromkeys(docList[0].keys(), 0)
          for word, val in idfDict.items():
              idfDict[word] = math.log10(N / (float(val) + 1))
          return(idfDict)
      #inputing our sentences in the log file
      idfs = computeIDF([wordDictA, wordDictB])
      print(idfs)
     {'21st': 0.3010299956639812, 'for': 0.3010299956639812, 'is':
     0.3010299956639812, 'of': 0.3010299956639812, 'sexiest': 0.3010299956639812,
     'data': 0.3010299956639812, 'learning': 0.3010299956639812, 'Data':
     0.3010299956639812, 'key': 0.3010299956639812, 'century': 0.3010299956639812,
     'machine': 0.3010299956639812, 'job': 0.3010299956639812, 'Science':
     0.3010299956639812, 'the': 0.3010299956639812, 'science': 0.3010299956639812}
[64]: def computeTFIDF(tfBow, idfs):
          tfidf = {}
          for word, val in tfBow.items():
              tfidf[word] = val*idfs[word]
          return(tfidf)
      #running our two sentences through the IDF:
      idfFirst =computeTFIDF(tfFirst, idfs)
      idfSecond = computeTFIDF(tfSecond, idfs)
      #putting it in a dataframe
      idf= pd.DataFrame([idfFirst, idfSecond])
      print(idf)
            21st
                                                               data learning \
                       for
                                            of
                                                  sexiest
     0 0.030103 0.000000 0.030103 0.030103 0.030103 0.000000 0.000000
     1 \quad 0.000000 \quad 0.037629 \quad 0.037629 \quad 0.000000 \quad 0.000000 \quad 0.037629 \quad 0.037629
            Data
                            century
                                       machine
                                                            Science
                                                                          the \
                       key
                                                      job
     0 0.030103 0.000000 0.030103 0.000000 0.030103 0.030103 0.060206
     1 0.000000 0.037629 0.000000 0.037629 0.000000 0.000000 0.037629
         science
     0.000000
     1 0.037629
[67]: #first step is to import the library
      from sklearn.feature_extraction.text import TfidfVectorizer
```

```
#for the sentence, make sure all words are lowercase or you will run
#into error. for simplicity, I just made the same sentence all
#lowercase

firstV= "Data Science is the sexiest job of the 21st century"
secondV= "machine learning is the key for data science"

#calling the TfidfVectorizer
vectorize= TfidfVectorizer()

#fitting the model and passing our sentences right away:
response= vectorize.fit_transform([firstV, secondV])
print(response)
```

(0, 2)	0.24342026926924518
(0, 10)	0.24342026926924518
(0, 4)	0.24342026926924518
(0, 12)	0.48684053853849035
(0, 11)	0.34211869506421816
(0, 5)	0.34211869506421816
(0, 9)	0.34211869506421816
(0, 0)	0.34211869506421816
(0, 1)	0.34211869506421816
(1, 2)	0.28986933576883284
(1, 10)	0.28986933576883284
(1, 4)	0.28986933576883284
(1, 12)	0.28986933576883284
(1, 8)	0.40740123733358447
(1, 7)	0.40740123733358447
(1, 6)	0.40740123733358447
(1, 3)	0.40740123733358447