```
In [3]: import nltk
In [4]: | nltk.download('punkt') # For tokenization
        nltk.download('stopwords') # For stopwords
        nltk.download('wordnet') # For Lemmatization
        nltk.download('averaged_perceptron_tagger') # For POS tagging
        [nltk_data] Downloading package punkt to
        [nltk data]
                        C:\Users\Welcome\AppData\Roaming\nltk_data...
        [nltk_data]
                      Package punkt is already up-to-date!
        [nltk data] Downloading package stopwords to
        [nltk_data]
                        C:\Users\Welcome\AppData\Roaming\nltk_data...
                      Package stopwords is already up-to-date!
        [nltk data]
        [nltk_data] Downloading package wordnet to
                        C:\Users\Welcome\AppData\Roaming\nltk_data...
        [nltk_data]
        [nltk_data]
                      Package wordnet is already up-to-date!
        [nltk data] Downloading package averaged perceptron tagger to
        [nltk_data]
                        C:\Users\Welcome\AppData\Roaming\nltk_data...
        [nltk_data]
                      Package averaged_perceptron_tagger is already up-to-
        [nltk_data]
Out[4]: True
         #Step 2: Initialize Text
In [5]:
         text = "Natural Language Processing helps computers understand human language.
In [7]: from nltk.tokenize import sent_tokenize
        text = "Natural Language Processing helps computers understand human language."
        # Sentence Tokenization
        tokenized_text = sent_tokenize(text)
        print(tokenized_text)
        ['Natural Language Processing helps computers understand human language.']
In [9]: from nltk.tokenize import word tokenize
        text = "Natural Language Processing helps computers understand human language."
        # Word Tokenization
        tokenized_word = word_tokenize(text)
        print(tokenized_word)
        ['Natural', 'Language', 'Processing', 'helps', 'computers', 'understand', 'hum
        an', 'language', '.']
```

```
In [10]:
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         import re
         # Define stop words
         stop_words = set(stopwords.words("english"))
         # Sample text
         text = "How to remove stop words with NLTK library in Python?"
         # Remove punctuation (keep only letters)
         text = re.sub('[^a-zA-Z]', ' ', text)
         # Tokenize text
         tokens = word_tokenize(text.lower())
         # Remove stopwords
         filtered_text = [w for w in tokens if w not in stop_words]
         # Output results
         print("Tokenized Sentence:", tokens)
         print("Filtered Sentence:", filtered_text)
         Tokenized Sentence: ['how', 'to', 'remove', 'stop', 'words', 'with', 'nltk',
         'library', 'in', 'python']
         Filtered Sentence: ['remove', 'stop', 'words', 'nltk', 'library', 'python']
In [11]: from nltk.stem import PorterStemmer
         # List of words to stem
         e_words = ["wait", "waiting", "waited", "waits"]
         # Initialize the stemmer
         ps = PorterStemmer()
         # Perform stemming
         for w in e words:
             rootWord = ps.stem(w)
             print(rootWord)
         wait
         wait
         wait
         wait
```

```
In [12]: from nltk.stem import WordNetLemmatizer
         import nltk
         # Initialize the Lemmatizer
         wordnet lemmatizer = WordNetLemmatizer()
         # Sample text
         text = "studies studying cries cry"
         # Tokenize the text
         tokenization = nltk.word_tokenize(text)
         # Perform Lemmatization
         for w in tokenization:
             print("Lemma for {}: {}".format(w, wordnet_lemmatizer.lemmatize(w)))
         Lemma for studies: study
         Lemma for studying: studying
         Lemma for cries: cry
         Lemma for cry: cry
In [13]: from nltk.tokenize import word tokenize
         from nltk import pos_tag
         # Sample text
         data = "The pink sweater fit her perfectly"
         # Tokenize the text
         words = word_tokenize(data)
         # Apply POS tagging
         pos_tags = pos_tag(words)
         # Print the results
         print(pos_tags)
         [('The', 'DT'), ('pink', 'NN'), ('sweater', 'NN'), ('fit', 'VBP'), ('her', 'PR
         P$'), ('perfectly', 'RB')]
In [14]: #Part 2: TF-IDF Representation of Document
In [16]: # Step 1: Import Required Libraries
         import pandas as pd
         from sklearn.feature_extraction.text import TfidfVectorizer
         import math
In [17]: # Step 2: Initialize the Documents
         documentA = "Jupiter is the largest Planet"
         documentB = "Mars is the fourth planet from the Sun"
```

```
# Step 3: Create Bag of Words (BoW) for Document A and B
In [19]:
         bagOfWordsA = documentA.split(' ')
         bagOfWordsB = documentB.split(' ')
In [20]:
          bagOfWordsA
Out[20]: ['Jupiter', 'is', 'the', 'largest', 'Planet']
In [21]: bagOfWordsB
Out[21]: ['Mars', 'is', 'the', 'fourth', 'planet', 'from', 'the', 'Sun']
In [22]:
          # Step 4: Create Collection of Unique Words from Document A and B
         uniqueWords = set(bagOfWordsA).union(set(bagOfWordsB))
         print(uniqueWords)
         {'from', 'is', 'Sun', 'largest', 'the', 'fourth', 'Mars', 'planet', 'Planet',
          'Jupiter'}
In [23]:
         # Step 5: Create a Dictionary of Words and Their Occurrence for Each Document
         numOfWordsA = dict.fromkeys(uniqueWords, 0)
         for word in bagOfWordsA:
             numOfWordsA[word] += 1
         numOfWordsB = dict.fromkeys(uniqueWords, 0)
         for word in bagOfWordsB:
             numOfWordsB[word] += 1
         # Print the word occurrence dictionaries
         print("Word Occurrences in Document A:", numOfWordsA)
         print("Word Occurrences in Document B:", numOfWordsB)
         Word Occurrences in Document A: {'from': 0, 'is': 1, 'Sun': 0, 'largest': 1,
         'the': 1, 'fourth': 0, 'Mars': 0, 'planet': 0, 'Planet': 1, 'Jupiter': 1}
         Word Occurrences in Document B: {'from': 1, 'is': 1, 'Sun': 1, 'largest': 0,
         'the': 2, 'fourth': 1, 'Mars': 1, 'planet': 1, 'Planet': 0, 'Jupiter': 0}
In [24]:
          numOfWordsA
Out[24]: {'from': 0,
           'is': 1,
           'Sun': 0,
           'largest': 1,
           'the': 1,
           'fourth': 0,
           'Mars': 0,
           'planet': 0,
           'Planet': 1,
           'Jupiter': 1}
```

```
In [25]:
         numOfWordsB
Out[25]: {'from': 1,
           'is': 1,
           'Sun': 1,
           'largest': 0,
           'the': 2,
           'fourth': 1,
           'Mars': 1,
           'planet': 1,
           'Planet': 0,
           'Jupiter': 0}
In [26]: # Step 6: Compute Term Frequency (TF)
         def computeTF(wordDict, bagOfWords):
             tfDict = {}
             bagOfWordsCount = len(bagOfWords)
             for word, count in wordDict.items():
                  tfDict[word] = count / float(bagOfWordsCount)
             return tfDict
         # Compute TF for both Document A and B
         tfA = computeTF(numOfWordsA, bagOfWordsA)
         tfB = computeTF(numOfWordsB, bagOfWordsB)
         # Print Term Frequency (TF) for each document
         print("Term Frequency for Document A:", tfA)
         print("Term Frequency for Document B:", tfB)
         Term Frequency for Document A: {'from': 0.0, 'is': 0.2, 'Sun': 0.0, 'largest':
         0.2, 'the': 0.2, 'fourth': 0.0, 'Mars': 0.0, 'planet': 0.0, 'Planet': 0.2, 'Ju
          piter': 0.2}
         Term Frequency for Document B: {'from': 0.125, 'is': 0.125, 'Sun': 0.125, 'lar
          gest': 0.0, 'the': 0.25, 'fourth': 0.125, 'Mars': 0.125, 'planet': 0.125, 'Pla
         net': 0.0, 'Jupiter': 0.0}
In [27]:
          tfA
Out[27]: {'from': 0.0,
           'is': 0.2,
           'Sun': 0.0,
           'largest': 0.2,
           'the': 0.2,
           'fourth': 0.0,
           'Mars': 0.0,
           'planet': 0.0,
           'Planet': 0.2,
           'Jupiter': 0.2}
```

```
In [28]:
          tfB
Out[28]: {'from': 0.125,
           'is': 0.125,
           'Sun': 0.125,
           'largest': 0.0,
           'the': 0.25,
           'fourth': 0.125,
           'Mars': 0.125,
           'planet': 0.125,
           'Planet': 0.0,
           'Jupiter': 0.0}
In [29]: # Step 7: Compute Inverse Document Frequency (IDF)
         def computeIDF(documents):
             N = len(documents)
             idfDict = dict.fromkeys(documents[0].keys(), 0)
             for document in documents:
                  for word, val in document.items():
                      if val > 0:
                          idfDict[word] += 1
             for word, val in idfDict.items():
                  idfDict[word] = math.log(N / float(val))
             return idfDict
         # Compute IDF for the documents
         idfs = computeIDF([numOfWordsA, numOfWordsB])
         # Print Inverse Document Frequency (IDF)
         print("Inverse Document Frequency (IDF):", idfs)
          Inverse Document Frequency (IDF): {'from': 0.6931471805599453, 'is': 0.0, 'Su
         n': 0.6931471805599453, 'largest': 0.6931471805599453, 'the': 0.0, 'fourth':
          0.6931471805599453, 'Mars': 0.6931471805599453, 'planet': 0.6931471805599453,
          'Planet': 0.6931471805599453, 'Jupiter': 0.6931471805599453}
In [30]:
          idfs
Out[30]: {'from': 0.6931471805599453,
           'is': 0.0,
           'Sun': 0.6931471805599453,
           'largest': 0.6931471805599453,
           'the': 0.0,
           'fourth': 0.6931471805599453,
           'Mars': 0.6931471805599453,
           'planet': 0.6931471805599453,
           'Planet': 0.6931471805599453,
           'Jupiter': 0.6931471805599453}
```

```
In [31]: |# Step 8: Compute TF-IDF
         def computeTFIDF(tfBagOfWords, idfs):
             tfidf = {}
             for word, val in tfBagOfWords.items():
                 tfidf[word] = val * idfs[word]
             return tfidf
         # Compute TF-IDF for both Document A and B
         tfidfA = computeTFIDF(tfA, idfs)
         tfidfB = computeTFIDF(tfB, idfs)
         # Create a DataFrame for visualization
         df = pd.DataFrame([tfidfA, tfidfB])
         # Print the DataFrame
         print(df)
                from
                               Sun
                                     largest the
                                                     fourth
                                                                         planet
                      is
                                                                 Mars
         0 0.000000 0.0 0.000000 0.138629 0.0 0.000000 0.000000 0.000000
         1 0.086643 0.0 0.086643 0.000000 0.0 0.086643 0.086643 0.086643
```

Planet Jupiter
0 0.138629 0.138629
1 0.000000 0.000000

In [ ]:

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BATCH:B3