```
In [10]: import nltk
         from nltk.tokenize import word_tokenize
         from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer, WordNetLemmatizer
         import contractions
         from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
         import re
In [4]: # Download necessary NLTK data files
         nltk.download('punkt')
         nltk.download('stopwords')
         nltk.download('wordnet')
         [nltk_data] Downloading package punkt to
         [nltk_data]
                         C:\Users\satch\AppData\Roaming\nltk_data...
         [nltk_data]
                       Package punkt is already up-to-date!
         [nltk_data] Downloading package stopwords to
         [nltk_data]
                         C:\Users\satch\AppData\Roaming\nltk_data...
         [nltk_data]
                       Package stopwords is already up-to-date!
         [nltk_data] Downloading package wordnet to
                        C:\Users\satch\AppData\Roaming\nltk_data...
         [nltk_data]
         [nltk_data]
                       Package wordnet is already up-to-date!
         True
Out[4]:
 In [8]: text = "Hey there! \square I can't believe it's already 2024. Did you see John's new blog post
In [9]: # Function to preprocess text
         def preprocess_text(text):
             # 1. Expand Contractions
             text = contractions.fix(text)
             # 2. Remove URLs and Emails
             text = re.sub(r'http\S+|www\S+|https\S+|mailto:\S+', '', text, flags=re.MULTILINE)
             text = re.sub(r'\S+@\S+', '', text)
             # 3. Remove special characters and emojis
             text = re.sub(r'[^a-zA-Z\s]', '', text)
             # 4. Tokenization
             words = word_tokenize(text)
             # 5. Lowercasing
             words = [word.lower() for word in words]
             # 6. Removing Punctuation
             words = [word for word in words if word.isalnum()]
             # 7. Removing Stop Words
             stop_words = set(stopwords.words('english'))
             filtered_words = [word for word in words if word not in stop_words]
             # 8. Stemming
             stemmer = PorterStemmer()
             stemmed_words = [stemmer.stem(word) for word in filtered_words]
             # 9. Lemmatization
             lemmatizer = WordNetLemmatizer()
             lemmatized_words = [lemmatizer.lemmatize(word) for word in filtered_words]
             return ' '.join(lemmatized_words)
         # Preprocess the text
```

```
preprocessed_text = preprocess_text(text)
print("Preprocessed Text:", preprocessed_text)
```

Preprocessed Text: hey believe already see john new blog postcheck also email mentioned something stemming lemmatizationinteresting stuff wayi attending ai conference nyc next month excited ai let u catch soon cheer john

```
In [11]: # 10. Vectorization
                    # Using CountVectorizer
                    count_vectorizer = CountVectorizer()
                    count_vector = count_vectorizer.fit_transform([preprocessed_text])
                    print("Count Vectorizer - Feature Names:", count_vectorizer.get_feature_names_out())
                    print("Count Vectorizer - Vectorized Text:", count_vector.toarray())
                    # Using TfidfVectorizer
                    tfidf_vectorizer = TfidfVectorizer()
                     tfidf_vector = tfidf_vectorizer.fit_transform([preprocessed_text])
                    print("TF-IDF Vectorizer - Feature Names:", tfidf_vectorizer.get_feature_names_out())
                    print("TF-IDF Vectorizer - Vectorized Text:", tfidf_vector.toarray())
                    Count Vectorizer - Feature Names: ['ai' 'already' 'also' 'attending' 'believe' 'blog' 'c
                    atch' 'cheer'
                       'conference' 'email' 'excited' 'hey' 'john' 'lemmatizationinteresting'
                      'let' 'mentioned' 'month' 'new' 'next' 'nyc' 'postcheck' 'see'
                      'something' 'soon' 'stemming' 'stuff' 'wayi']
                    1 1]]
                    TF-IDF Vectorizer - Feature Names: ['ai' 'already' 'also' 'attending' 'believe' 'blog'
                     'catch' 'cheer'
                      'conference' 'email' 'excited' 'hey' 'john' 'lemmatizationinteresting'
                      'let' 'mentioned' 'month' 'new' 'next' 'nyc' 'postcheck' 'see'
                      'something' 'soon' 'stemming' 'stuff' 'wayi']
                    TF-IDF Vectorizer - Vectorized Text: [[0.34815531 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407766 0.17407760 0.17407760 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740776 0.1740
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