1. Tokenization

Tokenization is the process of breaking text into individual words or sentences.

Explanation:

Downloading 'punkt': nltk.download('punkt') downloads the Punkt tokenizer models, which are pre-trained models for tokenizing text.

Tokenizing Text: word_tokenize(text) splits the text into individual words and punctuation marks.

2. Removing Stop Words

Stop words are common words that are often removed from text data.

```
from nltk.corpus import stopwords
# Ensure the necessary NLTK data files are downloaded
nltk.download('punkt')
nltk.download('stopwords')
# Load English stopwords
stop_words = set(stopwords.words("english"))
text = "Where are you I am here doing natural language processing tasks"
# Tokenize the text
tokens = nltk.word_tokenize(text)
# Filter out stopwords
filtered_text = [word for word in tokens if word.lower() not in stop_words]
print(filtered_text)
→ ['natural', 'language', 'processing', 'tasks']
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk_data] Package punkt is already up-to-date!
     [nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data]
                  Package stopwords is already up-to-date!
```

3. Stemming

Stemming is the process of reducing words to their base or root form.

```
from nltk.stem import PorterStemmer
stemmer=PorterStemmer()
words=["jumping","running","runs","played"]
stemmed_words=[stemmer.stem(word) for word in words]
print(stemmed_words)
```

```
    ['jump', 'run', 'run', 'play']
```

4. Lemmatization

Lemmatization is the process of reducing words to their base or dictionary form.

```
from nltk.stem import WordNetLemmatizer
nltk.download('wordnet')

lemmatizer = WordNetLemmatizer()
words = ["running", "jumps", "easily", "fairly"]
lemmatized_words = [lemmatizer.lemmatize(word) for word in words]
print(lemmatized_words)

Inltk_data] Downloading package wordnet to /root/nltk_data...
['running', 'jump', 'easily', 'fairly']
```

5. Part-of-Speech Tagging

POS tagging assigns parts of speech to each word in a sentence.

DT: Determiner (e.g., "the", "a") JJ: Adjective (e.g., "quick", "lazy") NN: Noun, singular or mass (e.g., "fox", "dog") VBZ: Verb, 3rd person singular present (e.g., "jumps") IN: Preposition or subordinating conjunction (e.g., "over") .: Punctuation mark (e.g., ".")

```
nltk.download('averaged_perceptron_tagger')
sentence = "The quick brown fox jumps over the lazy dog."
pos_tags = nltk.pos_tag(word_tokenize(sentence))
print(pos_tags)

inltk_data] Downloading package averaged_perceptron_tagger to
    [nltk_data] /root/nltk_data...
    [nltk_data] Unzipping taggers/averaged_perceptron_tagger.zip.
    [('The', 'DT'), ('quick', 'JJ'), ('brown', 'NN'), ('jumps', 'VBZ'), ('over', 'IN'), ('the', 'DT'), ('lazy)
```

6. Named Entity Recognition

NER identifies named entities in text.

```
#This imports the spaCy library and loads the English language model ("en_core_web_sm").

import spacy
nlp = spacy.load("en_core_web_sm")

text = "Apple is looking at buying U.K. startup for $1 billion"
doc = nlp(text)

#Iterates over the entities recognized in the processed document (doc) and prints each entity's text (ent.text) along with it for ent in doc.ents:
    print(ent.text, ent.label_)

Apple ORG
U.K. GPE
$1 billion MONEY
```

Output Explanation The output shows the recognized entities and their corresponding labels:

Apple: Recognized as an organization (ORG). U.K.: Recognized as a geopolitical entity (GPE). \$1 billion: Recognized as a monetary value (MONEY).

```
# 7. Sentence Tokenization
# Sentence tokenization splits text into sentences.

from nltk.tokenize import sent_tokenize

text = "Hello world! How are you today? Welcome to NLP."
sentences = sent_tokenize(text)
print(sentences)
```

```
# 8. Text Normalization
# Text normalization converts text to a standard format.
import re

text = "This is an example text with punctuation, numbers 123 and UPPERCASE letters."
normalized_text = re.sub(r'\d+', '', text).lower()
print(normalized_text)

this is an example text with punctuation, numbers and uppercase letters.
```

re.sub(r'\d+', ", text): Uses the re.sub() function to substitute (replace) all sequences of digits (\d+) in the text with an empty string ", effectively removing the digits. .lower(): Converts the resulting text to lowercase.

Text normalization is an important preprocessing step in natural language processing tasks. While this example focuses on removing digits and converting text to lowercase

```
#9. Spell Checking
#Spell checking corrects spelling errors in text.

from textblob import TextBlob

text = "I havv a spelking error."
blob = TextBlob(text)
print(blob.correct())

I have a speaking error.
```

Start coding or generate with AI.