2. NLP – Text wrangling and cleaning

Contents

1. Text wrangling and cleansing	2
2. Sentence Splitting	3
3. Tokenization	5
4. Word tokenization	6
5. Stemming	8
6. Lemmatization	10
7. Stop word removal	12

2. NLP – Text wrangling and cleaning

1. Text wrangling and cleansing

- ✓ The ultimate goal of doing text processing is to, machine can understand the text data.
- ✓ Text processing is very important as it helps us understand our data better and gain insights from it.
- ✓ Text processing normally involves the following steps,
 - Tokenization
 - Lemmatization
 - Stemming
 - Stop word removal

Program Downloading punkt Name demo1.py

import nltk

nltk.download("punkt")

Output

[nltk_data] Downloading package punkt to

[nltk data] C:\Users\admin\AppData\Roaming\nltk data...

[nltk data] Package punkt is already up-to-date!

- ✓ We have then downloaded the punkt, which is a Tokenizer for a text into a list of sentences.
- ✓ We can now do text wrangling.

2. Sentence Splitting

- ✓ Generally we used to understand sentence by reading words
- ✓ We will start reading sentence to understand paragraph.
- ✓ So, a paragraph we need to split it into a set of sentences.
- ✓ So, let's understand words, sentences and paragraph.
- ✓ This we can easily do by using nltk package

Program Name

Sentence splitting by using split() method

demo2.py

myString = "This is a paragraph. It should split at the end of sentence marker, such as a period. It can tell that the period in Mr.Daniel is not an end. Run it!, Hey How are you doing"

result = myString.split(".")
print(result)

Output

['This is a paragraph', ' It should split at the end of sentence marker, such as a period', ' It can tell that the period in Mr', 'Daniel is not an end', ' Run it!, Hey How are you doing']

Program Name

Sentence splitting

demo3.py

from nltk.tokenize import sent tokenize

myString = "This is a paragraph. It should split at the end of sentence marker, such as a period. It can tell that the period in Mr.Daniel is not an end. Run it!, Hey How are you doing"

tokenized_sentence = sent_tokenize(myString)
print(tokenized_sentence)

Output

['This is a paragraph.', 'It should split at the end of sentence marker, such as a period.', 'It can tell that the period in Mr.Daniel is not an end.', 'Run it!, Hey How are you doing']

Note

- ✓ We can understand from above output, the paragraph was split into the exact sentences.
- ✓ It was also able to tell the difference between a period that has been used to end a sentence from one used on the name Mr.Daniel also

3. Tokenization

- ✓ Tokenization refers to the process by which a large text is broken down into various pieces.
- ✓ In NLP, a token is the minimal piece of text that a machine can be able to understand.
- ✓ A text may be tokenized into words or sentences.
 - o In the case of sentence tokenization, every sentence will be identified as a token.
 - o In the case of word tokenization, every word will be identified as a token.

4. Word tokenization

- ✓ There are various ways through which tokenization can be done, but the most popular way of doing it is through word tokenization.
- ✓ What happens in word tokenization is that a large text is broken down into words and the words are used as the tokens.
 - o word tokenizer
 - sent_tokenizer
 - o punkt tokenizer
 - o Regexp_tokenizer
- ✓ To tokenize a text, we can still apply split() method from string

```
Program Word tokenization
Name demo4.py

myString = "These are sentences. Let us tokenize it! Run it!"

print(myString.split())

Output

['These', 'are', 'sentences.', 'Let', 'us', 'tokenize', 'it!', 'Run', 'it!']
```

```
Program
Name

Word tokenization with word_tokenize function
demo5.py

from nltk.tokenize import word_tokenize

myString = "These are sentences. Let us tokenize it! Run it!"

print(word_tokenize(myString))

Output

['These', 'are', 'sentences', '.', 'Let', 'us', 'tokenize', 'it', '!', 'Run', 'it', '!']
```

Program Name	Word tokenization with regexp_tokenize function demo6.py
	from nltk.tokenize import regexp_tokenize
	myString = "These are sentences. Let us tokenize it! Run it!"
	<pre>print(regexp_tokenize(myString, pattern="\w+"))</pre>
Output	
	['These', 'are', 'sentences', 'Let', 'us', 'tokenize', 'it', 'Run', 'it']

Program Name	Capture the digit from sentence demo7.py
	from nltk.tokenize import regexp_tokenize
	myString = "These are 3 sentences. Let us tokenize them! Run the code!"
	<pre>print(regexp_tokenize(myString, pattern="\d+"))</pre>
Output	
	['3']

5. Stemming

- ✓ This is another step in text wrangling.
- ✓ From the name, you can get the meaning, cutting down a token to its root stem.
- ✓ Stemming algorithm works by cutting the suffix from the word.
- ✓ In simple, it cuts either the beginning or end of the word.
- ✓ Consider the word "cutting".
- ✓ This word can be broken down to its root, which is "cut".

```
Program Stemming
Name demo8.py

from nltk.stem import PorterStemmer

porter = PorterStemmer()
print(porter.stem("cutting"))

Output

cut
```

```
Program Stemming demo9.py

from nltk.stem import PorterStemmer

e_words= ["wait", "waiting", "waited", "waits"]

ps = PorterStemmer()

for w in e_words:
    rootWord = ps.stem(w)
    print(rootWord)

Output

wait
wait
wait
wait
wait
wait
wait
```

Program Name

Stemming demo10.py

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)))

Output

Stemming for studies is studi Stemming for studying is studi Stemming for cries is cri Stemming for cry is cri

- ✓ Stemming is good for its simplicity when it comes to dealing with NLP problems.
- ✓ However, when more complex stemming is needed, stemming is not the best option, but we have lemmatization.

6. Lemmatization

- ✓ Lemmatization is a more advanced compared to stemming.
- ✓ It follow specific rules to get results.
- ✓ It understand the context, parts of the speech to understand the root of the word

Program Lemmatization
Name demo11.py

from nltk.stem import WordNetLemmatizer

wl = WordNetLemmatizer()

print("rocks :", wl.lemmatize("rocks"))
print("corpora :", wl.lemmatize("corpora"))
print("better :", wl.lemmatize("better", pos = "a"))

Output

rocks : rock
corpora : corpus
better : good

Program Name Lemmatization demo12.py

import nltk

from nltk.stem import WordNetLemmatizer

wl = WordNetLemmatizer()

text = "studies studying cries cry"

tokens = nltk.word_tokenize(text)

for word in tokens:

print("Lemmatization for {} is {}".format(word, wl.lemmatize(word)))

Output

Lemmatization for studies is study Lemmatization for studying is studying Lemmatization for cries is cry Lemmatization for cry is cry

7. Stop word removal

- ✓ Some common words that are present in text but do not contribute in the meaning of a sentence.
- ✓ Such words are not at all important for the purpose of information retrieval or natural language processing.
- ✓ The most common stopwords are 'a', 'in' 'the' etc
- ✓ The good news is that nltk comes with a list of stop words and in different languages.

Program Name

list of the stopwords

demo13.py

import nltk

from nltk.corpus import stopwords

print(stopwords.words('english'))

Output

['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'such', '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', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't"]

Program list of the languages

Name demo13.py

from nltk.corpus import stopwords

langs = stopwords.fileids()

print(len(langs))

Output

29

Program Lemmatization
Name demo14.py

import nltk

from nltk.corpus import stopwords

nltk.download('stopwords')

mylist = stopwords.words('english')

line = "This is really good, how are you doing"

postPa = [word for word in line.split()]

print(postPa)

Output

['This', 'is', 'really', 'good,', 'how', 'are', 'you', 'doing']

Program Name	Lemmatization demo15.py
	import nltk from nltk.corpus import stopwords
	nltk.download('stopwords')
	mylist = stopwords.words('english')
	line = "This is really good, how are you doing"
	postPa = [word for word in line.split() if word not in mylist]
	print(postPa)
Output	['This', 'really', 'good,']