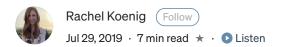


Get started



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NLP for Beginners: Cleaning & Preprocessing Text Data

NLP is short for Natural Language Processing. As you probably know, computers are not as great at understanding words as they are numbers. This is all changing though as advances in NLP are happening everyday. The fact that devices like Apple's Siri and Amazon's Alexa can (usually) comprehend when we ask the weather, for directions, or to play a certain genre of music are all examples of NLP. The spam filter in your email and the spellcheck you've used since you learned to type in elementary school are some other basic examples of when your computer is understanding language.

As a data scientist, we may use NLP for sentiment analysis (classifying words to have positive or negative











For this example, we're examining a dataset of Amazon products/reviews which can be found and downloaded for free on <u>data.world</u>. I'll be using Python in Jupyter notebook.

Here are the imports used:

```
1
   #Import pandas
 2
   import pandas as pd
 3
   #Import Natual Language Toolkit
 4
   import nltk
 5
   #Import Beautiful Soup
 6
   from bs4 import BeautifulSoup
 7
   #Import string for list of punctuation
 8
   import string
   # Import the stop word list
 9
10
   from nltk.corpus import stopwords
   # Import Tokenizer
11
   from nltk.tokenize import RegexpTokenizer
12
   #Import Lemmatizer
13
   from nltk.stem import WordNetLemmatizer
14
   # Import stemmer.
15
   from nltk.stem.porter import PorterStemmer
16
```

(You may need to run nltk.download() in a cell if you've never previously used it.)

Read in csv file, create DataFrame & check shape. We are starting out with 10,000 rows and 17 columns. Each row is a different product on Amazon.

```
1 df = pd.read_csv('amazon_co-ecommerce_sample.csv')
1 df.shape
(10000, 17)
```

I conducted some basic data cleaning that I won't go into detail about now, but you can read my post about EDA <u>here</u> if you want some tips.

In order to make the dataset more manageable for this example, I first dropped columns with too many nulls and then dropped any remaining rows with null values. I changed the <code>number_of_reviews</code> column type from object to integer and then created a new DataFrame using only the rows with no more than 1 review. My new shape is 3,705









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The following workflow is what I was taught to use and like using, but the steps are just general suggestions to get you started. Usually I have to modify and/or expand depending on the text format.

- 1. Remove HTML
- 2. Tokenization + Remove punctuation
- 3. Remove stop words
- 4. Lemmatization or Stemming

While cleaning this data I ran into a problem I had not encountered before, and learned a cool new trick from geeksforgeeks.org to split a string from one column into multiple columns either on spaces or specified characters.

The column I am most interested in is <code>customer_reviews</code>, however, upon taking a closer look, it currently has the review title, rating, review date, customer name, and review all in one cell separated by //.

```
1 review_df['customer_reviews'][3]
"I love it // 5.0 // 22 July 2013 // By\n \n Lilla Lukacs\n \n on 22 July 2013 // I lo
ve it. Perfect with the earlier ordered locomotive.Again: I would recommend it to the masters
of the topic. It's not just a toy."
```

Pandas .str.split method can be applied to a Series. First parameter is the repeated part of the string you want to split on, n= maximum number of separations and expand=True will split up the sections into new columns. I set the 4 new columns equal to a new variable called reviews.

```
#separate review column by // breaks
2
    reviews = review_df['customer_reviews'].str.split("//", n=4, expand=True)
 3
    reviews.head()
                       0
                  I love it 5.0
                                                                                         I love it. Perfect with the earlier ordered I...
                               22 July 2013
                                                By\n \n Lilla Lukacs\n \n on 22 July 2...
3
8
               Five Stars 5.0 23 Dec. 2015
                                                By\n \n Mr. K. Waller\n \n on 23 Dec. ... Xmas box for grandson,am sure it will be great
10
    steaming good engine! 5.0 11 Jan. 2016
                                               By\n \n Jeremy Joslin\n \n on 11 Jan. ...
                                                                                               Grandson loved it. Hauls good load
                               12 Nov. 2014
                                             By\n \n alain HENAUX\n \n on 12 Nov. 2...
                                                                                                   Very good product. Thank you.
               Five Stars
                         5.0
11
12
               Five Stars 5.0
                                2 Oct. 2015
                                              By\n \n Mark Ritzmann\n \n on 2 Oct. 2...
                                                                                                                      very good
```

Then you can rename the new 0, 1, 2, 3, 4 columns in the original reviews_df and drop the original messy column.

```
1    review_df["review_title"] = reviews[0]
2    review_df['rating'] = reviews[1]
3    review_df['review_date'] = reviews[2]
```



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review	customer_name	review_date	rating	review_title
I love it. Perfect with the earlier ordered I	Lilla Lukacs	22 July 2013	5.0	I love it
Xmas box for grandson,am sure it will be great	Mr. K. Waller	23 Dec. 2015	5.0	Five Stars
Grandson loved it. Hauls good load	Jeremy Joslin	11 Jan. 2016	5.0	steaming good engine!
Very good product. Thank you.	alain HENAUX	12 Nov. 2014	5.0	Five Stars
very good	Mark Ritzmann	2 Oct. 2015	5.0	Five Stars

I ran the same method over the new <code>customer_name</code> column to split on the <code>\n \n</code> and then dropped the first and last columns to leave just the actual customer name. There is a lot more we could do here if this were a longer article! Right off the bat, I can see the names and dates could still use some cleaning to put them in a uniform format.

Removing HTML is a step I did not do this time, however, if data is coming from a web scrape, it is a good idea to start with that. This is the function I would have used.

```
def remove_html(text):
    soup = BeautifulSoup(text, 'lxml')
    html_free = soup.get_text()
    return html_free
```









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go, but I wanted to show the break down and make them easier to customize.

Remove punctuation:

One way of doing this is by looping through the Series with list comprehension and keeping everything that is not in string.punctuation, a list of all punctuation we imported at the beginning with import string.

```
def remove_punctuation(text):
    no_punct = "".join([c for c in text if c not in string.punctuation])
    return no_punct
```

"".join will join the list of letters back together as words where there are no spaces.

```
review_df['review'] = review_df['review'].apply(lambda x: remove_punctuation
review_df['review'].head()

I love it Perfect with the earlier ordered lo...
Xmas box for grandsonam sure it will be great
Grandson loved it Hauls good load
Very good product Thank you
very good
```

If you scroll up you can see where this text previously had commas, periods, etc.

However, as you can see in the second line of output above, this method does not account for user typos. Customer had typed "grandson,am" which then became one word "grandsonam" once the comma was removed. I still think this is handy to know in case you ever need it though.

Tokenize:

This breaks up the strings into a list of words or pieces based on a specified pattern using Regular Expressions aka RegEx. The pattern I chose to use this time (r'\w') also removes punctuation and is a better option for this data in particular. We can also add.lower() in the lambda function to make everything lowercase.

```
#Instantiate Tokenizer
    tokenizer = RegexpTokenizer(r'\w+')
    review df['review'] = review df['review'].apply(lambda x: tokenizer.tokenize(x.lower()))
    review df['review'].head(20)
3
      [i, love, it, perfect, with, the, earlier, ord...
      [xmas, box, for, grandson, am, sure, it, will,...
8
10
               [grandson, loved, it, hauls, good, load]
11
                      [very, good, product, thank, you]
12
                                            [very, good]
      [great, quality, lots, of, detail, run, nic, a...
14
```



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Some other examples of RegEx are:

```
'\w+|\s|^2 = splits up by spaces or by periods that are not attached to a digit
```

Remove stop words:

We imported a list of the most frequently used words from the NL Toolkit at the beginning with from nltk.corpus import stopwords. You can run stopwords.word(insert language) to get a full list for every language. There are 179 English words, including 'i', 'me', 'my', 'myself', 'we', 'you', 'he', 'his', for example. We usually want to remove these because they have low predictive power. There are occasions when you may want to keep them though. Such as, if your corpus is very small and removing stop words would decrease the total number of words by a large percent.

```
def remove stopwords(text):
 2
        words = [w for w in text if w not in stopwords.words('english')]
 3
        return words
    review df['review'] = review df['review'].apply(lambda x : remove stopwords(x))
 2
    review df['review'].head(10)
3
      [love, perfect, earlier, ordered, locomotive, ...
8
                     [xmas, box, grandson, sure, great]
10
                    [grandson, loved, hauls, good, load]
11
                                  [good, product, thank]
12
                                                   [good]
      [great, quality, lots, detail, run, nic, smoot...
14
15
      [fantastic, little, set, beautifully, made, de...
18
                                 [lovely, quality, kato]
      [second, bachmann, powerhauls, coming, take, r...
19
20
      [excellent, ho, scale, diorama, piece, compris...
```

Stemming & Lemmatizing:

Both tools shorten words back to their root form. Stemming is a little more aggressive. It cuts off prefixes and/or endings of words based on common ones. It can sometimes be helpful, but not always because often times the new word is so much a root that it loses its actual meaning. Lemmatizing, on the other hand, maps common words into one base. Unlike stemming though, it always still returns a proper word that can be found in the dictionary. I like to compare the two to see which one works better for what I need. I usually prefer Lemmatizer, but surprisingly, this time, Stemming seemed to have more of an affect.







^{&#}x27;\s+', gaps=True = grabs everything except spaces as a token

 $^{&#}x27;[A-Z]\w+' = only words that begin with a capital letter.$



15 18

19

20

Open in app

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```
1
    # Instantiate lemmatizer
 2
    lemmatizer = WordNetLemmatizer()
 3
 4
    def word lemmatizer(text):
 5
        lem text = [lemmatizer.lemmatize(i) for i in text]
        return lem text
 6
    review df['review'].apply(lambda x: word lemmatizer(x))
3
        [love, perfect, earlier, ordered, locomotive, ...
                        [xmas, box, grandson, sure, great]
8
10
                       [grandson, loved, haul, good, load]
11
                                    [good, product, thank]
12
                                                     [good]
        [great, quality, lot, detail, run, nic, smooth...
14
```

Lemmatizer: can barely even see a difference

[lovely, quality, kato]

You see more of a difference with Stemmer so I will keep that one in place. Since this is the final step, I added "".join() to the function to join the lists of words back together.

[fantastic, little, set, beautifully, made, de...

[second, bachmann, powerhauls, coming, take, r...

[excellent, ho, scale, diorama, piece, compris...

```
#Instantiate Stemmer
 2
    stemmer = PorterStemmer()
 1
    def word stemmer(text):
 2
        stem text = " ".join([stemmer.stem(i) for i in text])
 3
        return stem text
    review df['review'] = review df['review'].apply(lambda x: word stemmer(x))
    review df['review']
3
        love perfect earlier order locomot would recom...
8
                              xma box grandson sure great
10
                              grandson love haul good load
11
                                        good product thank
12
                                                       good
        great qualiti lot detail run nic smooth great ...
14
               fantast littl set beauti made decent price
15
                                         love qualiti kato
18
19
        second bachmann powerhaul come take real train...
        excel ho scale diorama piec compris white encl...
20
```



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