

NATURAL LANGUAGE PROCESSING (NLP)

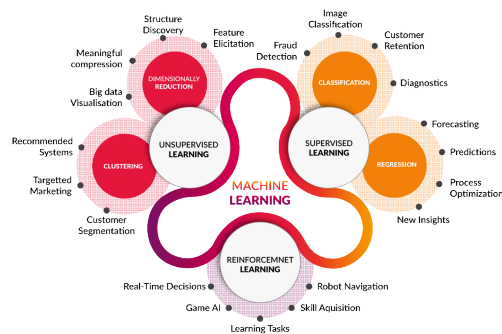
Machine Learning
Dr. Adnan Abid

Natural Language Processing - NLP

Natural-language processing (NLP) is an area of computer science and artificial intelligence concerned with the interactions between computers and human (natural) languages. NLP is used to apply Machine Learning models to text and language.

Example use:

Teach machines to understand what is said in spoken and written word is the focus of Natural Language Processing. Whenever you dictate something into your iPhone / Android device that is then converted to text, that's an NLP algorithm in action.



NLP History

The history of natural-language processing generally started in the 1950s, although work can be found from earlier periods. In 1950, Alan Turing published an article titled "Computing Machinery and Intelligence" which proposed what is now called the Turing test as a criterion of intelligence.

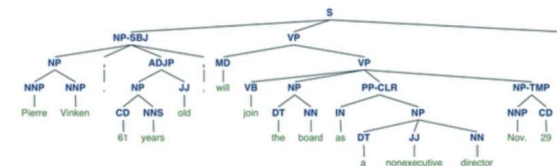
Up to the 1980s, most natural-language processing systems were based on complex sets of hand-written rules. Starting in the late 1980s, however, there was a revolution in natural-language processing with the introduction of machine learning algorithms for language processing.

https://en.wikipedia.org/wiki/Natural-language_processing

NLP Uses

- Sentiment analysis. Identifying the mood or subjective opinions within large amounts of text, including average sentiment and opinion mining.
- Use it to predict the genre of the book.
- Question Answering
- Use NLP to build a machine translator or a speech recognition system
- Document Summarization

NATURAL LANGUAGE PROCESSING - NLP



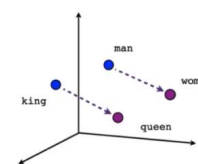
<https://www.nltk.org/>

Natural Language Processing - NLP

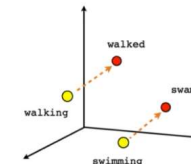
Main NLP library examples:

Natural Language Toolkit - NLTK
 SpaCy
 Stanford NLP
 OpenNLP

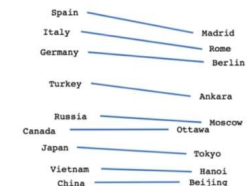
Natural Language Processing - NLP



Male-Female



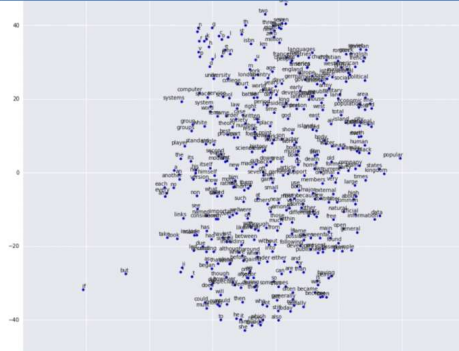
Verb tense



Country-Capital

<https://www.tensorflow.org/tutorials/word2vec>

Natural Language Processing - NLP



Natural Language Processing

In this part, you will understand and learn how to:

1. Clean texts to prepare them for the Machine Learning models,
2. Create a Bag of Words model,
3. Apply Machine Learning models onto this Bag of Words model.

NLP - Bag of Words

Very popular NLP model - It is a model used to preprocess the texts to classify before fitting the classification algorithms on the observations containing the texts.

It involves two things:

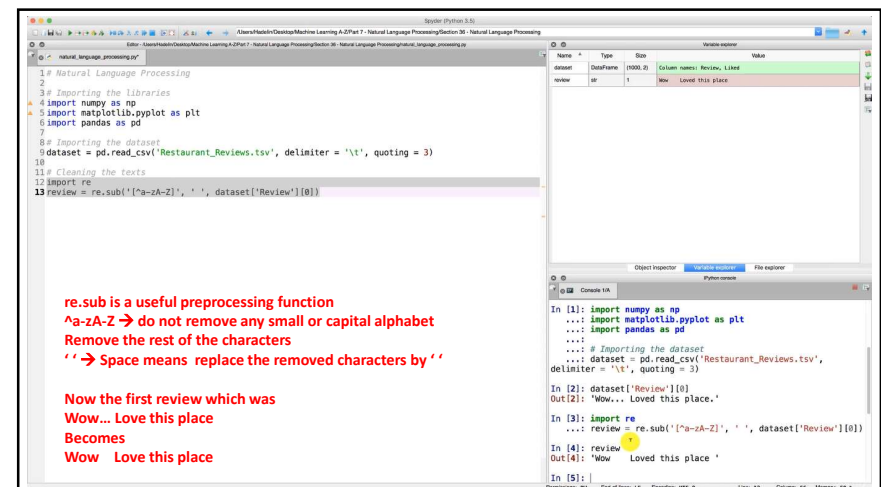
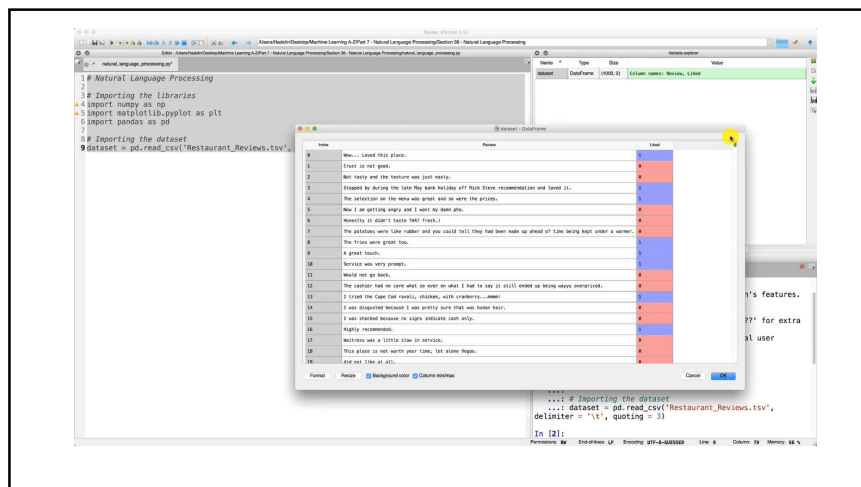
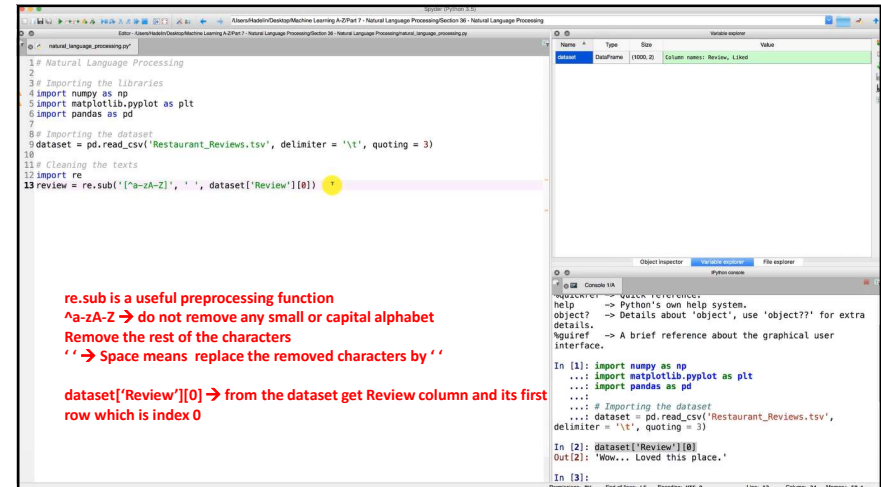
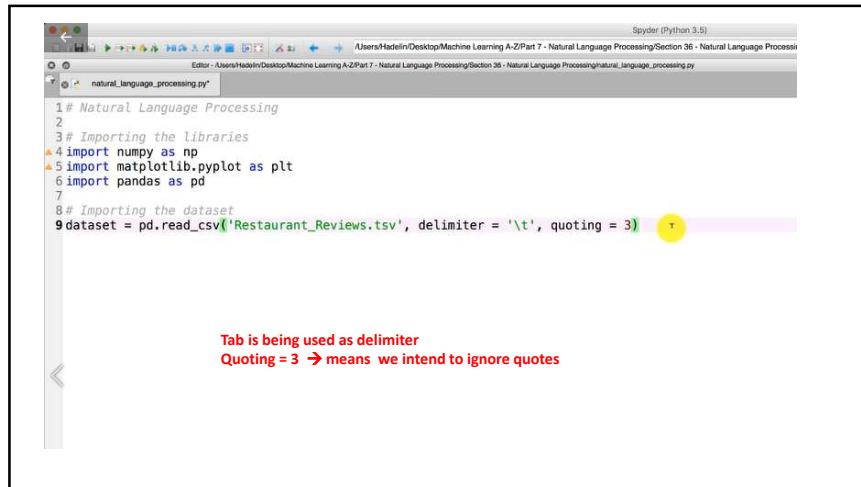
1. A vocabulary of known words.
2. A measure of the presence of known words.

```

Review Liked
Wow... Loved this place.,1
Crust is not good.,0
Not tasty and the texture was just nasty.,0
Stopped by during the late May bank holiday off Rick Steve recommendation and loved it.,1
The selection on the menu was great and so were the prices.,1
Now I am getting angry and I want my damn pho.,0
Honestly it didn't taste THAT fresh.,0
The potatoes were like rubber and you could tell they had been made up ahead of time being
kept under a warmer.,0
The fries were great too.,1
A great touch.,1
Service was very prompt.,1
Would not go back.,0
The cashier had no care what so ever on what I had to say it still ended up being waxyx
overpriced.,0
I tried the Cape Cod rayall, chicken, with cranberry...mmm!,1
I was disgusted because I was pretty sure that was human hair.,0
I was shocked because no signs indicate cash only.,0
Highly recommended.,1
Waitress was a little slow in service.,0
This place is not worth your time, let alone Vegas.,0
did not like at all.,0
The Burritos Blah!,0
The food, amazing.,1
Service is also cute.,1
I could care less... The interior is just beautiful.,1
So they performed.,1
That's right...the red velvet cake...ohhh this stuff is so good.,1
- They never brought a salad we asked for.,0
  
```

CSV OR TSV?

Commas may be a part of the review comment... which may be misleading
Reviews do not contain any tabs, so TSV is suitable.



re.sub is a useful preprocessing function
 ^a-zA-Z → do not remove any small or capital alphabet
 Remove the rest of the characters
 '' → Space means replace the removed characters by ''

Now the first review which was
 Wow... Love this place
 Becomes
 Wow Love this place

Now convert all reviews into lower case...
 First review becomes
 wow love this place

The next step is to apply Stemming
 Convert the words into their root form
 Computer → compute
 Loved → Love
 Talking → Talk
 It reduces the overall vocabulary and helps reducing the size of the sparse matrix.

Porter Stemmer is one of widely used stemmer and nltk uses it too.
 Apply stemming after removing the stopwords so that the remain words are stemmed.
 Now the comment [wow loved place] becomes [wow love place]

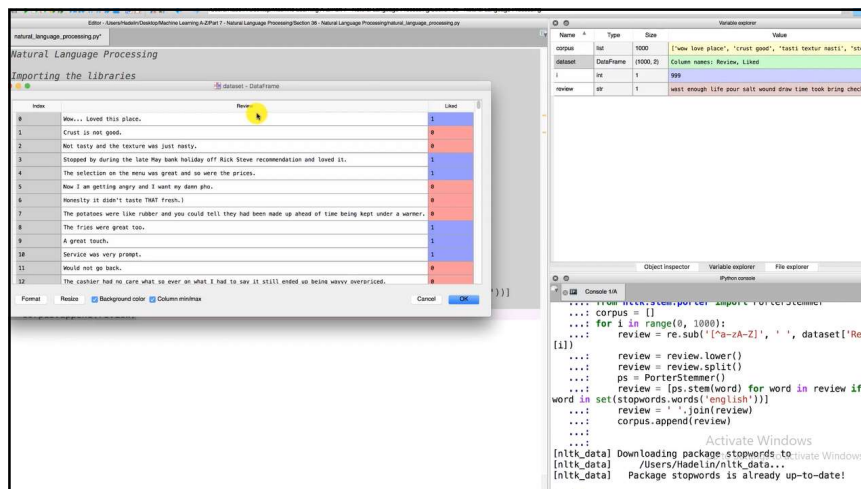
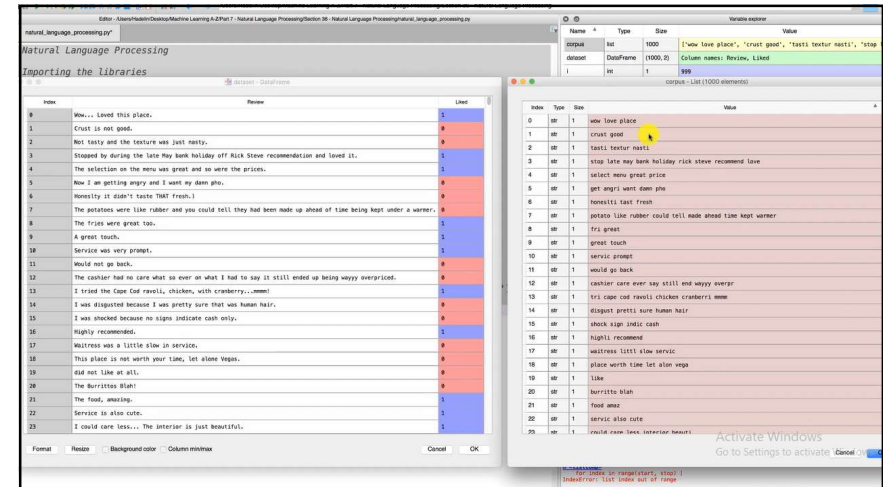
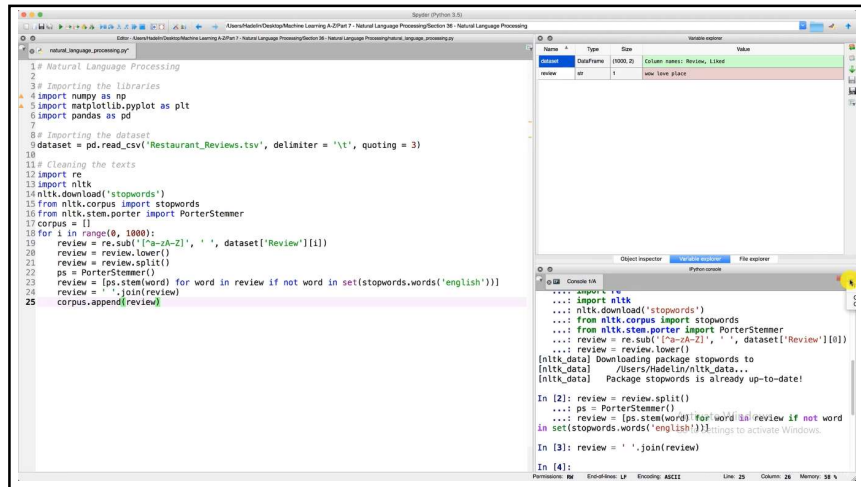
Now we need to remove STOP WORDS
 These are the words which do not have any meaning towards the positivity or negativity of the comment, but they are there to complete the structure of the sentences e.g. this, that, of, on etc.

NLTK provides a list of stop words.
 Download that list and import stopwords package from nltk.corpus

Apply a loop on all the words of the review comment, which have been split, and remove the ones which are STOP WORDS.

Use set function to check for stopwords as it works faster

The next step is to join all the words again to convert the preprocessed review into a single string by concatenating them with a ' '



Text Cleaning

- Load Text
- CSV vs TSV
- Convert into lower case
- Split the review comment/doc into words
- Remove stop words
- Stemming
- Combining all reviews/documents

```

1 # Natural Language Processing
2
3 # Importing the libraries
4 import numpy as np
5 import matplotlib.pyplot as plt
6 import pandas as pd
7
8 # Importing the dataset
9 dataset = pd.read_csv('Restaurant_Reviews.tsv', delimiter = '\t', quoting = 3)
10
11 # Cleaning the texts
12 import re
13 import nltk
14 nltk.download('stopwords')
15 from nltk.corpus import stopwords
16 from nltk.stem.porter import PorterStemmer
17 corpus = []
18 for i in range(0, 1000):
19     review = re.sub('[^a-zA-Z]', ' ', dataset['Review'][i])
20     review = review.lower()
21     ps = PorterStemmer()
22     review = [ps.stem(word) for word in review if not word in set(stopwords.words('english'))]
23     review = ' '.join(review)
24     corpus.append(review)
25
26 # Creating the Bag of words model
27 from sklearn.feature_extraction.text import CountVectorizer
28 cv = CountVectorizer()
29 X = cv.fit_transform(corpus).toarray()

```

converts the corpus into matrix

THE BAG OF WORDS model

Sparse matrix with one row for each review/document

***The CountVectorizer class can perform many of the preprocessing steps but it is better to explicitly perform those steps (A) for teaching purpose (B) we have the liberty to control many things e.g. in case of web scrapping the HTML tags also contain a-zA-Z letters...

```

30 cv = CountVectorizer(max_features = 1500)
31 X = cv.fit_transform(corpus).toarray()
32 y = dataset.iloc[:, 1].values

```

Set y to be the last column of the dataset.

Our data set has 2 columns 0 and 1.

Thus $y = \text{dataset.iloc[:, 1]}$ → all rows and 2nd column

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30 cv = CountVectorizer(max_features = 1500)
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Max_features = 1500 removes the words appearing in very few reviews. E.g. any proper nouns person's name etc.

It is kind of **dimensional reduction**

There are other methods too

Principal Component Analysis

Singular Value Decomposition

```

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Now, it has become a conventional ML problem

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26 # Splitting the dataset into the Training set and Test set
27 from sklearn.cross_validation import train_test_split
28 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, random_state=0)
29
30 # Fitting Naive Bayes to the Training set
31 from sklearn.naive_bayes import GaussianNB
32 classifier = GaussianNB()
33 classifier.fit(X_train, y_train)
34
35 # Predicting the Test set results
36 y_pred = classifier.predict(X_test)
37
38 # Making the Confusion Matrix
39 from sklearn.metrics import confusion_matrix
40 cm = confusion_matrix(y_test, y_pred)
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28 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=0)
29
30 # Feature Scaling
31 from sklearn.preprocessing import StandardScaler
32 sc = StandardScaler()
33 X_train = sc.fit_transform(X_train)
34 X_test = sc.transform(X_test)
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No need for feature scaling.
The values mostly 0 or 1, and very few 2s or 3s.

Also we cannot visualize data in 1500 dimensions, so we shall rely only on confusion matrix

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