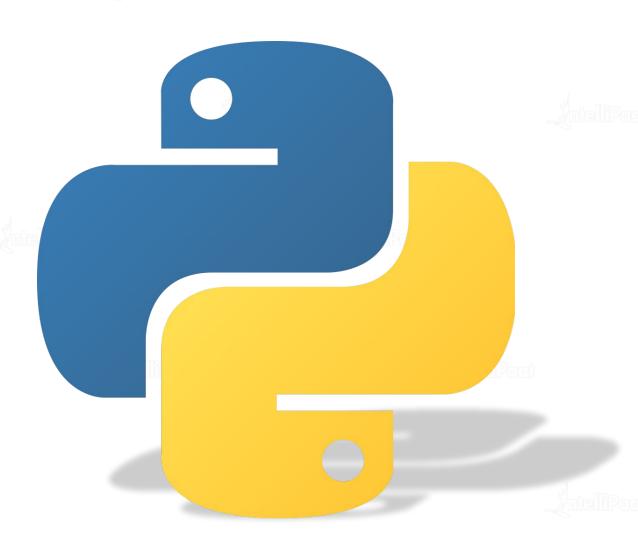


# Natural Language Processing





## Agenda

O1 What is NLP?

02 Text Analytics

o3 Sentiment Analysis



## What is Natural Language Processing?

#### What is Natural Language Processing?

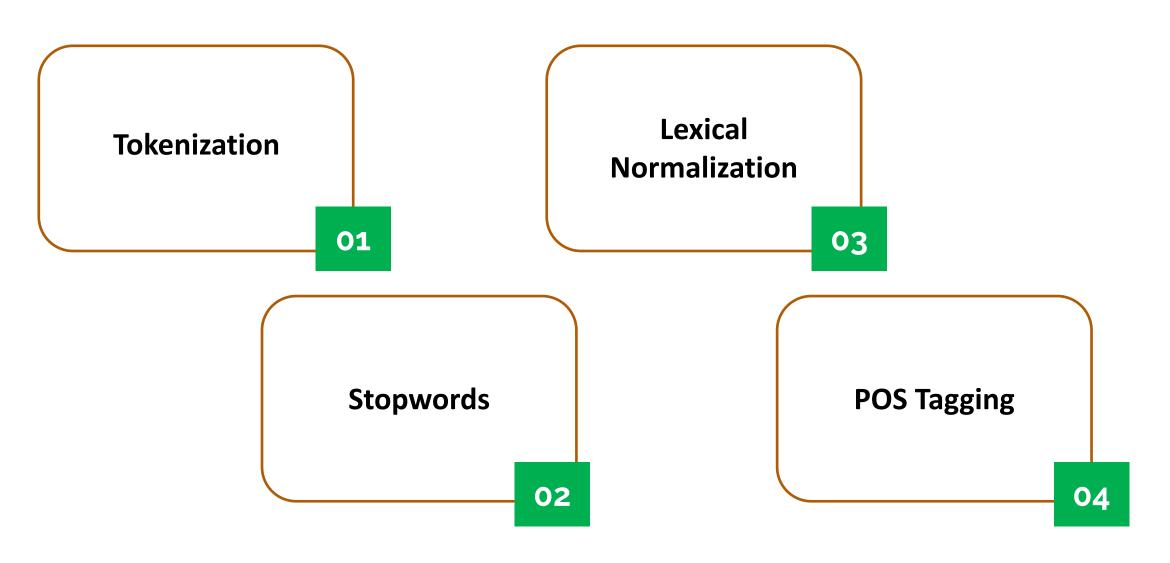




Natural language is how humans communicate, and with the generation of tons and tons of textual data. To analyze and process the data has become necessary for business outcomes.









Tokenization 01

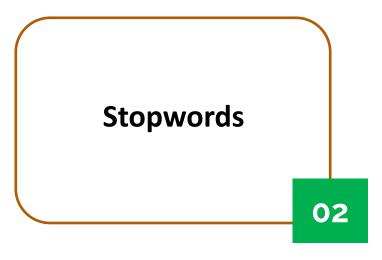
```
a = "Python is Popular. And, Python is easy to learn."
token = sent_tokenize(a)
print(token)

['Python is Popular.', 'And, Python is easy to learn.']
```

Tokenization is the process of breaking down the text into small tokens.

The processing time increases with tokenization and helps in faster computation of text analytics.





```
from nltk.corpus import stopwords
stop_words=set(stopwords.words("french"))
print(stop_words)

{'sommes', 'y', 'eurent', 'eux', 'me', 'ses',
```

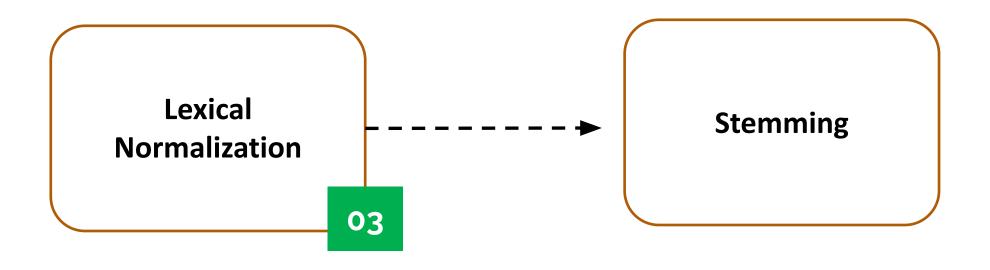
Stop words are the noise in the textual data, that must be removed from the data.



Lexical Normalization

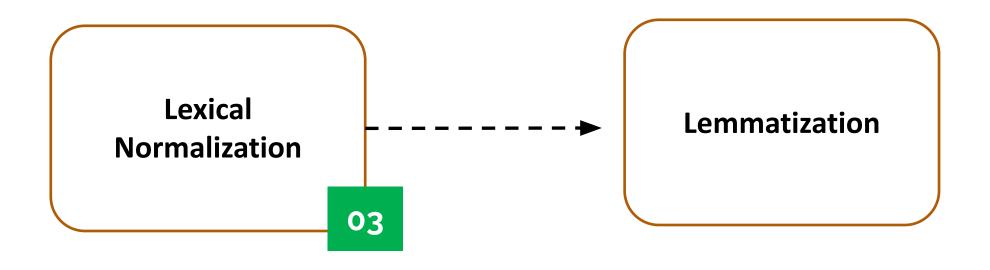
Lexical normalization is breaking down the tokens into their base forms, There are two types of lexical normalization – Stemming and Lemmatization.





Stemming is a linguistic normalization procedure in which the word or the token is reduced to its root word and removes the derivational affixes.





Lemmatization reduces words to their base word, which is linguistically correct lemmas. It transforms root word with the use of vocabulary and morphological analysis. For Example – Flying becomes fly.

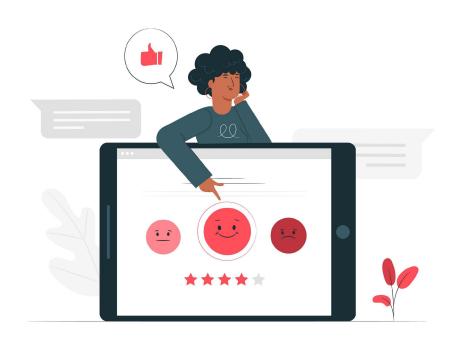


POS Tagging
04

POS tagging or the Part of Speech tagging is identifying the grammatical group for the given word. Like a noun, pronoun, etc.

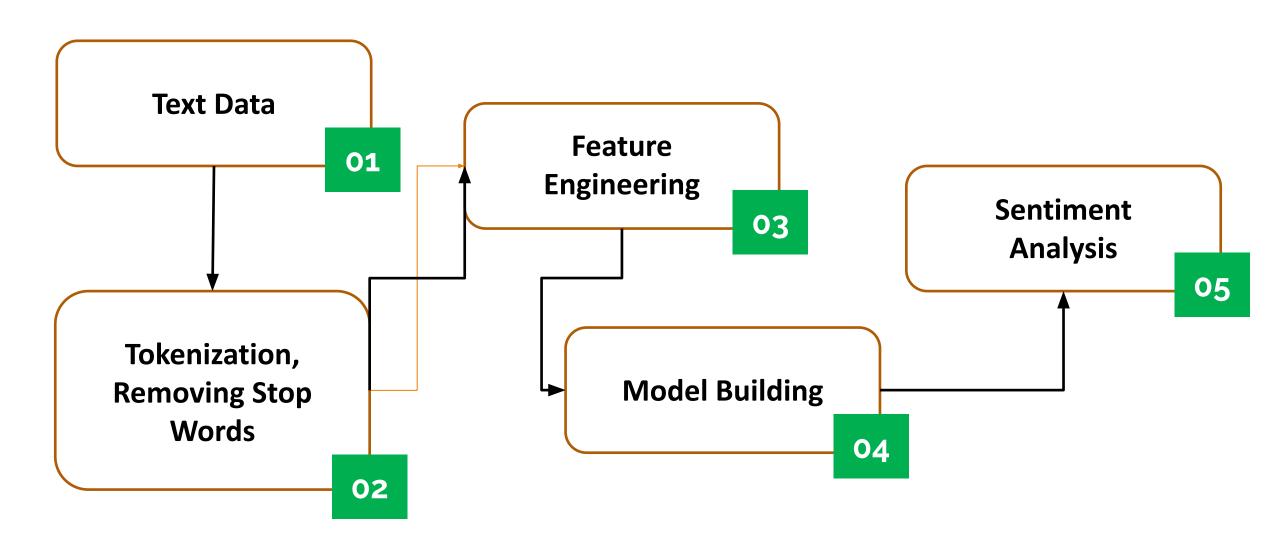






Sentiment Analysis is the predictive modeling technique in which a model is trained on sentiment labels, often generated by NLTK functions and then are used to predict the sentiments associated with the texts.







Text Data
01

The sentiment analysis lifecycle starts from gathering the textual data and transforming the data removing the outliers and null values before the further computation.



Tokenization,
Removing Stop
Words

02

Tokenization is initially done to simplify the computation, and stop words are removed to remove the noise from the data.



Feature Engineering

03

Feature engineering is done on the textual data by Bag-of-words and TF-IFD, that can be used to train the model.



Model Building

04

After the feature engineering has been finished, the model is trained on the transformed data, and we can use them to predict the sentiments on the test sets/validation sets, and unseen data as well.



Sentiment Analysis

05

For data, that may not have labels, we generate the sentiments associated with the texts based on the intensity using NLTK, and train the model with the exact same approach as we train any other model.