## Working of Bag-of-Words Model

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Date of Submission: 15th October 2020

in partial fulfillment for the award of the degree

of

Bachelor of Technology

In

Information Technology

At



Department of Information Technology
National Institute of Technology Karnataka, Surathkal
October 2020

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## Chapter 1

## Results

### 1.1 Introduction

Whenever an algorithm in NLP is applied, it works on a collection of numbers. Thus, we cannot feed a set of words, paragraphs, text documents etc. to the algorithm as input directly, and must be converted into a representation in the form of numbers. Hence, the Bag-of-Words model is used to convert any text into its namesake, which keeps a count of the total occurrences of the most frequently used words.

The working of the model can be visualized using a table, with a count of each word displayed.

Before displaying the working of the model, some preprocessing of the text is required.

## 1.2 Preprocessing

## 1.2.1 Stopword removal

Stopwords are words in a language that do not add any meaning or semantic information to a sentence. These can be removed while keeping the meaning of the sentence intact. For example, the English words

- and
- the
- but

are stopwords. Since the selection of stopwords is highly subjective, the 'stopwords' package in the nltk library is downloaded for use and applied on a paragraph as demonstrated below:



#### 1.2.2 Tokenization

Tokenization refers to the splitting up of words in a text into smaller words (for English), or even the creation of new words (in other languages). The NLTK library offers several ways of tokenization, such as:

- Sentence Tokenization: Splits text into sentences
- Word Tokenization: Splits text into words
- WordPunctuationTokenizer: Splits text into words as well as punctuation marks

Their working is demonstrated below:



Figure 1.2: Tokenization

## 1.2.3 Stemming

Stemming refers to reduction of words into their root form. For example, the words do, doing, done all have the same root form 'do'. While processing text containing these words using NLP algorithms, they should be treated the same as they have the same meaning. Thus, it is imperative to reduce all words in a text to their root forms. The NLTK library again offers various stemmers such as:

- Porter Stemmer
- Lancaster Stemmer
- Snowball Stemmer (Supports multiple languages)

Their working is demonstrated below:



Figure 1.3: Stemming

#### 1.2.4 Lemmatization

Lemmatization, similar to stemming, is the reduction of words into their base form. This base form is achieved by removing morphological differences and utilizing the vocabulary of the language. For example, the words good, better, best could all have the same base form 'good'. The same reasons for utilizing stemming apply to lemmatization, and stemming may not reduce words to forms that lemmatization does, as stemming uses the root form which is always part of the word. NLTK offers the WordNet Lemmatizer, whose working is demonstrated below:

```
After Lemmitiation using bordestimestimes descending a loss hown sizely a Alexa, is a virtual system at reconstant descend by Apazon, first used in the Section Etch sent speaker developed by Apazon ish 186. It is a speaker of use to interaction, must in Slapack, and and to do like, scattage alone, a reconsing speakers, pulping audicobooks, and providing weather, traffic, sports, and other real-time information, such a news.[3] Alexa can also control several senart device using itself a a home automation system. Users are able to extend the Alexa capability by installing "skills" (auditional functionality developed by third-p arty vendors, in words between Stemming and Lemmatization respectively:

Dat beginning and Lemmatization respectively:

Dat beginning and semantization respectively:

Dat sharing in gare

Saw Saw

feet foot

stripe stripe strips
```

Figure 1.4: Lemmatization

## 1.3 Vocabulary Construction

The BoW algorithm builds a model using a document-term matrix, constructed by considering the number of occurrences of each word in a document. Using this, each document can be represented as a weighted combination of various words in it. By setting a threshold and choosing words that carry more meaning, a histogram of all the words present in the documents in the form of a feature vector can be made.

Using the Count Vectorizer in the Scikit-learn library, a BoW model is constructed as demonstrated below:



Figure 1.5: BoW construction

## 1.4 N-gram modelling

An n-gram is a contiguous sequence of n items from a given sample of text or speech. These items can either be phonemes, syllables, letters, words, or base pairs depending on the application. N-grams are generally collected from a corpus containing text or speech. N-grams are used to generate language models that can predict which word comes next given a history of words.

Bigram, N-gram and Everygram modelling as part of the NLTK library is demonstrated below:



Figure 1.6: N-gram modelling

## 1.5 Text Corpora

Practical work in the field of NLP generally uses large bodies of linguistic corpora or data. Corpora are designed to carefully balance material across one or more genres.

Some of the corpora present in the NLTK library are (along with analysis demonstrated):

## 1.5.1 Gutenberg Corpus

NLTK includes a small selection of texts from the Project Gutenberg electronic text archive, which itself contains about 25,000 electronic books including established literature for free.

```
The files in the Gutenberg Corpus are: austen-eema.txt, austen-persuasion.txt, austen-persuasion.txt, austen-persuasion.txt, austen-persuasion.txt, austen-persuasion.txt, austen-persuasion.txt, desterton-ball.tx, chesterton-brown.txt, chesterton-brown.txt, chesterton-brown.txt, chesterton-brown.txt, chesterton-brown.txt, chesterton-brown.txt, shakespeare-hanlet.txt, shakespeare-nacheth.txt, whitman-leaves.txt Mumber of sounds (including punctuation) in Persuasion of the sounds are: [, Poens, by, William, Blake, 1789, ], SONGS, OF, INNOCENCE, AND, OF, EXPERIENCE, and, THE, BODK, of, THEL, SONGS, OF
The mumber of sentences in Hanlet by Shakespeare are: 3106
Sone of the words are: [, Poens, by, William, Blake, 1789, ], SONGS, OF, INNOCENCE, AND, OF, EXPERIENCE, and, THE, BODK, of, THEL, SONGS, OF
The mumber of sentences in Hanlet by Shakespeare are: 3106
Sone of Hanlet by William Shakespeare 1899 ]
Actus Primary
Act
```

Figure 1.7: Gutenberg Corpus

## 1.5.2 Brown Corpus

NLTK also includes the Brown corpus, which was the worlds-first million word electronic corpus in English, created in 1961 at Brown university. The corpus contains material from over 500 sources, each categorized by genre such as news, editorial etc.

```
The categories in the Brown corpus are: adventure, belles_lettres, editorial, fiction, government, hobbles, humor, learned, lore, mystery, mews, religion, reviews, romance, science_fiction

The number of words in the Humor category are: 21695
Some of the words are: It, was, among, these, that, Hinkle, identified, a, photograph, of
The number of sentences in the Fiction category are: 4249
The sentences are:
The number of sentences in the Fiction category are: 4249
The tributers are:
The sentence in the Fiction category is of length: 102
The centence is: Ne could no longer build anything, whether a private residence in his Pennsylvania county or a church in Brazil, without it being obvious that he had done it, and while here and there he was
taken to task for again developing the same airly technique, they were such fanctful and sometimes even playful buildings that the public felt assured by its sense of recognition after a time, a quality of authentic uniqueness about them, which, once established by an artist as his private vision, is no longer disputable as to its other values.

A conditional Frequency Distance between some words in the lens category:

Can-34 could 87 may; 39 might; 38 must; 33 vill; 389

A Conditional Frequency Distance tablation between one genres and words in the Brown Corpus:

Per section of the s
```

Figure 1.8: Brown Corpus