# ELL786 Assignment-3

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Abstract—This document describes our submission for Assignment 3 for the ELL786 (Multimedia Systems) course. Various techniques and methods which were used to develop a text based search engine, image based search engine and multi-modal search engine have been described.

Index Terms—TF-IDF, Word2Vec, K-means, SIFT, BoVW

#### I. TEXT BASED SEARCH ENGINE

#### A. Introduction

The text based search engine takes input as a word and use *Algo 1: TF-IDF* and *Algo 2: Word2Vec* to output the articles which are closest to the input word.

We used **Wikipedia** package of python to get the wikipedia article (using *wikipedia.page(articleName)*) page from the article name.

# B. Approach

#### 1) Preprocessing

The data (words in the documents) were preprocessed before using the algorithms by the following ways:

#### a) Lowercase

Lowercase is used to describe the shorter, smaller versions of letters (like w), called lowercase letters, as opposed to the bigger, taller versions (like W). This was done through python in-built function, "lower()".

# b) Removing Stop words

A stop word is a commonly used word (such as "the") that a search engine has been programmed to ignore, both when indexing entries for searching and when retrieving them as the result of a search query. Stop words are any word in a stop list which are filtered out before or after processing of natural language data. Ex- 'this', 'the', 'and', 'but', etc. We made an array of all the stop words and removed it whenever we encountered any stop words in the documents.

### c) Removing Contractions

A contraction is a shortened form of a word (or group of words) that omits certain letters or sounds. In most contractions, an apostrophe represents the missing letters. The most common contractions are made up of verbs, auxiliaries, or modals attached to other words. Ex- "He'd": He would, "ain't": "are not", etc. We made a dictionary for the contractions corresponding to their original/full forms and updated the contracting words to their full original forms.

#### d) Stemming

Stemming is the process of reducing a word to its word stem that affixes to suffixes and prefixes. It is just like cutting down the branches of a tree to its stems. For example, the stem of the words eating, eats, eaten is eat. Also, the stem of the word studies is studi. This was done using **Porter stemming algorithm** from python **NLTK** library.

#### e) Lemmatization

Lemmatization usually refers to doing things properly with the use of a vocabulary and morphological analysis of words, normally aiming to remove inflectional endings only and to return the base or dictionary form of a word, which is known as the lemma. For example, the root word of studies is study. This was done using **NLTK Lemmatization** with NLTK Tokenization.

#### 2) Algorithm I: TF-IDF

TF-IDF stands for "Term Frequency — Inverse Document Frequency". This is a technique to quantify words in a set of documents. We generally compute a score for each word to signify its importance in the document and corpus. This method is a widely used technique in Information Retrieval and Text Mining.

# $sklearn.feature\_extraction.text.TfidfVectorizer.fit\_transform$

was used to transform documents to document-term matrix. And then using *cosine similarity* between the input word and the term matrix we get the results.

# 3) Algorithm II: Word2vec

The *Word2vec* algorithm uses a neural network model to learn word associations from a large corpus of text. *Word2vec* takes as its input a large corpus of text and produces a vector space with each unique word in the corpus being assigned a corresponding vector in the space. Word vectors are positioned in the vector space such that words that share common contexts in the corpus are located close to one another in the space.

Word2vec was imported from **gensim.models** and using Word2vec() a model is trained (example: word2vec\_model = Word2Vec(train\_data, size=self.embeddingSize, window=5, sg=1). We can get the embedding using word2vec\_model.wv, all the embedding were stored in a list (docEmbeddings). And then using *cosine similarity* between the input word and the docEmbeddings we get the results.

#### C. Results

Fig. 1. Result of text based search

# II. IMAGE BASED SEARCH ENGINE

#### A. Introduction

The Image based search engine takes an image as input and show the images which are closest visually to the input image. The images corresponding to a word were downloaded using **icrawler**.

# B. Approach

# 1) Preprocessing

We resized image to max (1000\*1000) pixels maintaining the aspect ratio. Firstly, we calculated the aspect ratio

$$\mathbf{aspectratio} = \frac{\mathbf{height}}{\mathbf{width}}$$

from the original images. For images whose atleast one dimension (width or height) was greater than 1000 pixels. We

make the larger dimension 1000 pixels while calculate the smaller dimension using the aspect ratio. However, images smaller than (1000\*1000) were left as it is.

Also, the images were changed to grayscale as grayscale image is a one layer image from 0-255 whereas the RGB have three different layer image, while the features remain intact.

#### 2) Method

Then, using the module icrawler, we downloaded the images. Using SIFT algorithm in openCV library we extract the keypoint descriptors of each image. We train a KNN on all these descriptors. To get vectors for an image(in our dataset as well as an query image) we create a histogram from the KNN predicted classes of each of the descriptors in the image. We also do TF-IDF re-weighting. Finally we return the closest images to a query image based on consine similarity.

#### C. Results

On running the command

```
python3 Text_Search/main.py
```

we get the following results:

Fig. 2. Result of Image based search

Enter surl to search from image url, sfile to search from file, exit to exit...

sfile

Enter file path...

Dualipa.jpg

Search took 0.14409589767456055 s...

[1/5] ("Cricket", 'images/Cricket.jpg')

Enter next to show next result, anything else to search again

next

[2/5] ("Dualipa", 'images/Dualipa.jpg')

Forter next to show next result, anything else to search again

Forter next to show next result, anything else to search again

Fig. 3. Result of Image based search

#### A. Introduction

Given a word or image as input, Multimodal Search Engine return the closest words and images.

# B. Approach

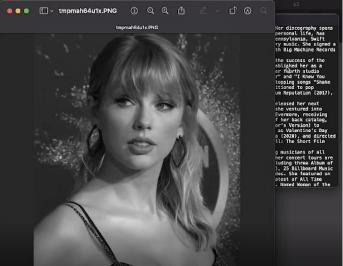
User can select whether he wants to run text-search or image-search, If the text-search is selected then closest words are returned using text based search engine, If image-search is selected then closest images are returned using image based search engine.

# C. Results

Fig. 4. Result of Multi modal search engine

```
Enter text-search to search from text, img-search to search from image text-search
Enter text search query game
Search took 8.8438058492889484 s...
Press Enter to Show next result or type exit to search again next
Press Enter to Show next result or type exit to search again exit
Enter text-search to search from text, img-search to search from image ing-search
Enter surl to search from image url, sfile to search from file sfile
Enter file path...
tay.jpg
Search took 8.87867764286744529 s...
Press Enter to Show next result or type exit to search again
```

Fig. 5. Result of Multi modal search engine



IV. DEMO

We wrote and ran the code in VS Code on Macbook M1, which has 8-core CPU with 4 performance cores and 4 efficiency cores, 8-core GPU, 16-core Neural Engine along with 256GB SSD.

Click here to watch Demo videos of the working code.

[1] Wikipedia: Popular Pages

[2] Python: Icrawler

[3] OpenCV: SIFT

[4] Word2Vec: Gensim[5] sklearn: TF-IDF