Project:	Chinese Word Segmentation System
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Team Leader:	Wu Zhengke
Team Member(s):	Wu Zhengke
Date:	2017/12/31

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Brief Introduction

This project is the assignment of Introduction to Computer Science in Dec. 2017.

In this project, a Chinese Word Segmentation System mainly base on HMM, Viterbi Algorithm and Maximum Matching Algorithm is built.

In addition, a local website is developed with Flask and Twitter Bootstrap as the User Interface.

1. Prototype System Introduction

1.1 Functions

This system mainly has the following five functions:

a. Training with given data
b. Cutting Chinese text into sentences
c. Segmenting Chinese sentences or text into words
d. User settings to correct the segmentation
e. A local website as User Interface

1.2 Running Environment

Windows 10 or others (not tested yet)
Python 3.6.0 (with packages installed in requirements.txt)

1.3 Developing Environment

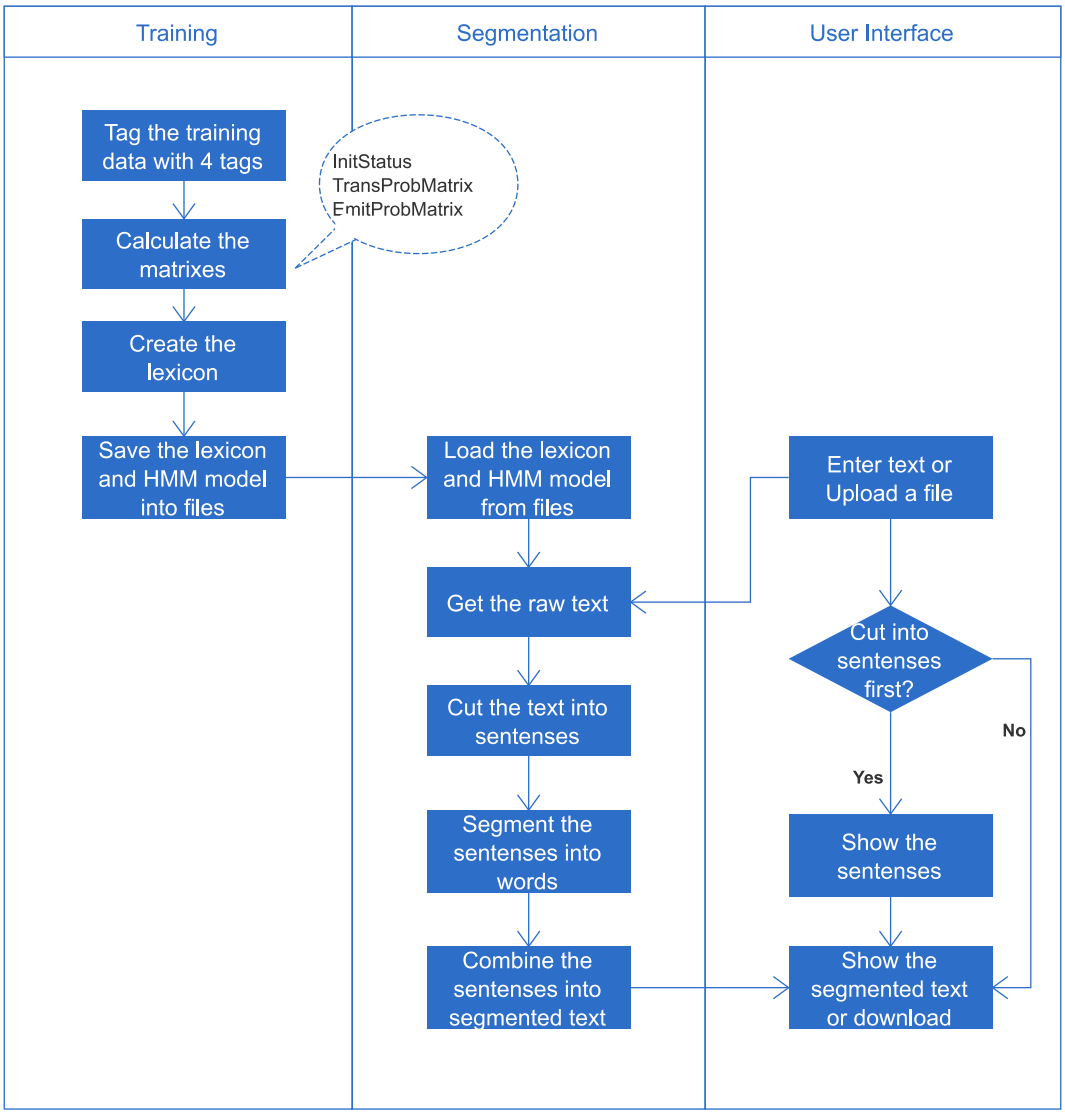
Windows 10
Python 3.6.0 (with packages installed in requirements.txt)
Sublime Text 3

2. Task Allocation

System Overall Design:	Wu Zhengke
Algorithm:	Wu Zhengke
User Interface:	Wu Zhengke
Debug and Test:	Wu Zhengke
Report:	Wu Zhengke

3. System Architecture

3.1 Overall Flowchart



As illustrated in the above flowchart, the system is mainly composed of three components: Training, Segmentation and User Interface.

3.2 Training

The program opens the given training files and tags every character in them with 4 tags: B, E, S, M. Then according to the tags, it counts the occurrence of tags and calculates 3 matrixes: InitStatus, TransProbMatrix and EmitProbMatrix. (This part will be further explained in Section 4.)

After all these works done, the HMM model, i.e. the 3 matrixes, are saved into files.

In addition, the program creates a lexicon with all the words in the given training data and save it into files.

3.3 Segmentation

First, the program loads the trained HMM model and the lexicon into memory from files. When dealing with the given raw text, it cuts the text into sentences using the punctuations as separator and then segments every sentence before combining them together.

(This part will also be further explained in Section 4.)

3.4 User Interface

There are two ways for users to input text: entering text directly or uploading a file.

Then the program transfers the raw text to the part of Segmentation and show the segmented text. And if the user wants

to cut the text into sentences first, it will also show the sentences.

For convenience, users can download the result.

4. Algorithm Description

This program is mainly based on the HMM (Hidden Markov Model). Before using this model, the program tags every character in the training corpus with 4 tags: B, M, E, S, which stand for Begin, Middle, End, Single respectively. For instance, the sentence can be tagged in this way:

今天 是 礼拜天 。
B E S B M E S

Then the program counts all the occurrence of these tags in the corpus and calculates the following 3 matrixes:

a. InitStatus: The possibility of each status of the first character in a sentence.
   For example:
   Initstatus['B']=0.6
   Initstatus['S']=0.4
   InitStatus['M']=0
   InitStatus['E']=0

b. TransProbMatrix: The possibility of transformation from Status A to Status B
   For example:
   TransProbMatrix['B']['M']=0.4
   TransProbMatrix['B']['E']=0.6
   TransProbMatrix['B']['S']=0
   TransProbMatrix['B']['B']=0

c. EmitProbMatirx: The possibility of a certain character in a certain status

   For example:

   EmitProbMatrix[‘S’][‘我’]=0.0001

   EmitProbMatrix[‘S’][‘们’]=0

When segmenting, the program first cuts the text into sentences according to punctuations. And to preserve the punctuations when splitting, it’s better to use the sub() in the re module(regular expression) rather than split().

Then the program uses these three matrixes to determine the most possible status of each character in a sentence. To specify, that’s Viterbi Algorithm.

From the beginning character of a sentence, the program calculates the most possible trace of status. For the first character, its status traces are “B”,“M”,“E”,“S” and corresponding possibilities are InitStatus[status] * EmitProbMatrix[status][character].

And for every following character, its possible status are ‘B’, ‘M’, ‘E’, ‘S’,and correspondingly, the trace of a certain status is the trace of the previous character and its status that maximizes:

Possibility[previous_character][previous_status] *
TransProbMatrix[previous_status][status] *
EmitProbMatrix[status][character],

appending the certain status.

Manipulate in this way until the end of a sentence, then the program gets the possible traces of the last character and the maximum of them is the most possible status trace of the whole sentence. And the last step is to segment the sentence according to the status trace and combine all the sentences together into the text.

But after some experiments, the pure HMM model doesn't work well enough (You may see the result in Section 5.2) as it is weak in dealing with IV (In Vocabulary) words. To offset this deficiency, this program combines the HMM with Maximum Matching Algorithm to handle IV words.

As for Maximum Matching Algorithm, it's relatively easy. For a given sentence, it searches from the beginning of the sentence to match the longest word in the lexicon.

If it fails to match any word that begins with a character, then it adds this character to a string, which will be handled with HMM later.

In this way, this program is better at dealing with both IV and OOV words.

5. Demo and Testing Result

5.1 Demo

The demo can be got by forking from the Github repository[1] or unpacking the ZIP file attached.

But before running it, you need to install all the packages that are listed in requirements.txt. (A simple way to do this is to run *pip install -r requirements.txt*)

To run the demo, open the folder FlaskUI/ and run the command:

python FlaskUI.py runserver

Then you can visit *127.0.0.1:5000(or localhost:5000)* to use this demo.

In the index page, you can enter text, choose whether to cut it into sentences first and then click Segment to see the result.(See Figure 1, Figure 2,Figure 3.)

[1] https://github.com/keithnull/ChineseWordSegmentationSystem

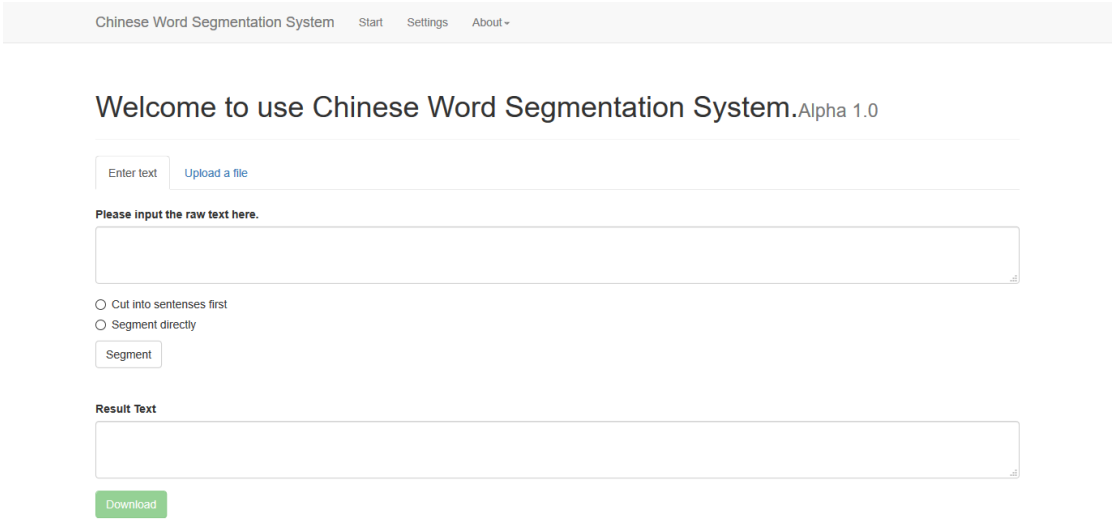


Figure 1

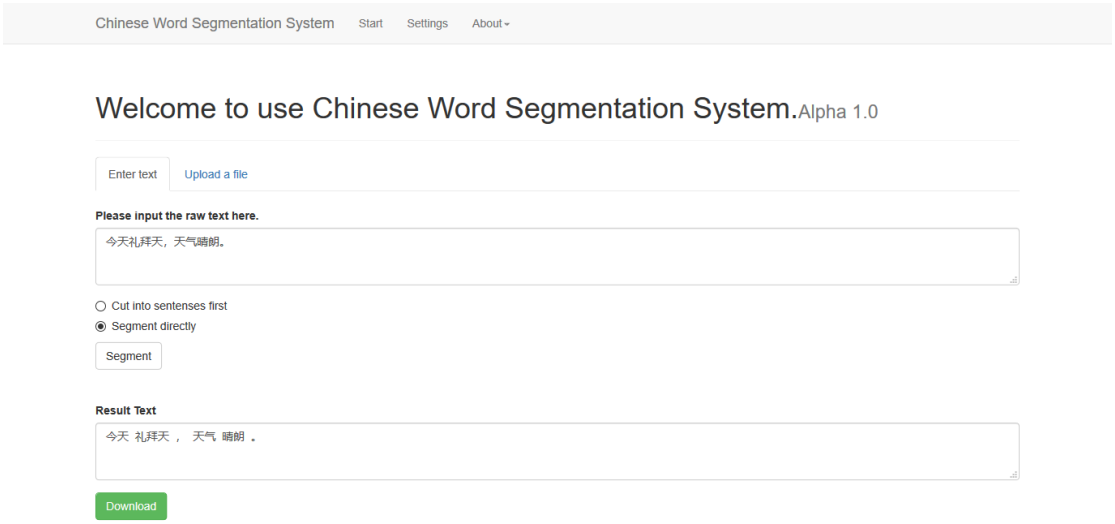


Figure 2

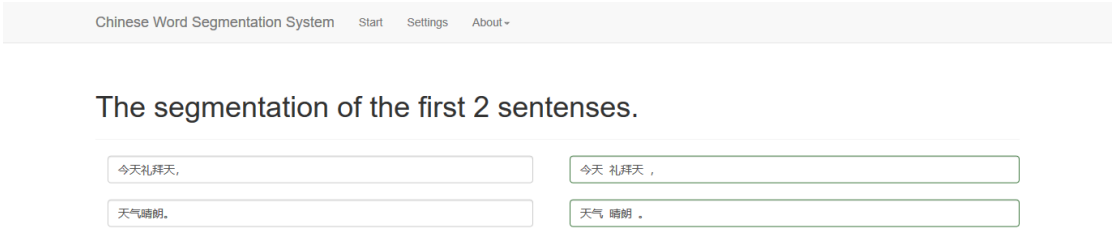


Figure 3

Also, you can upload a file. But note that it must be encoded in UTF-8. (See Figure 4, Figure 5.)

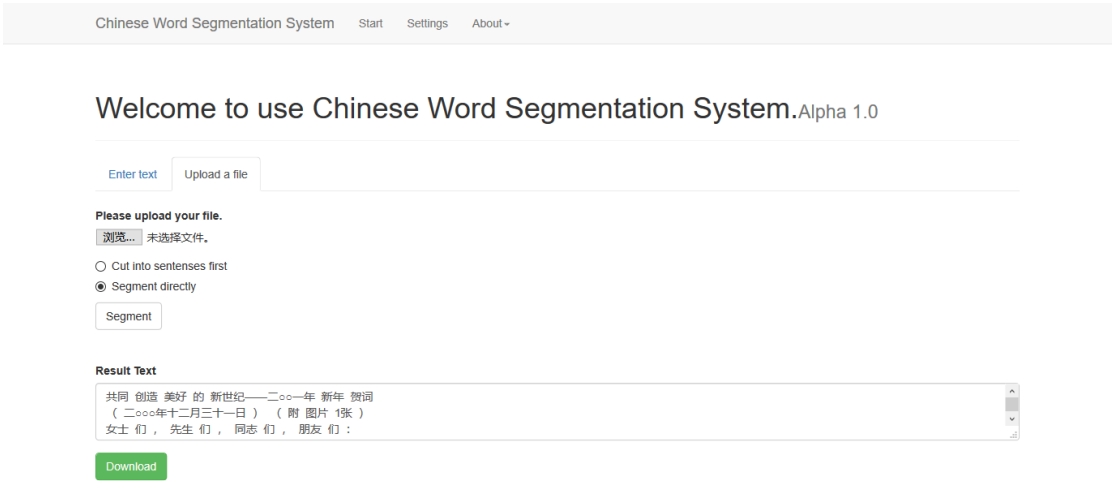


Figure 4

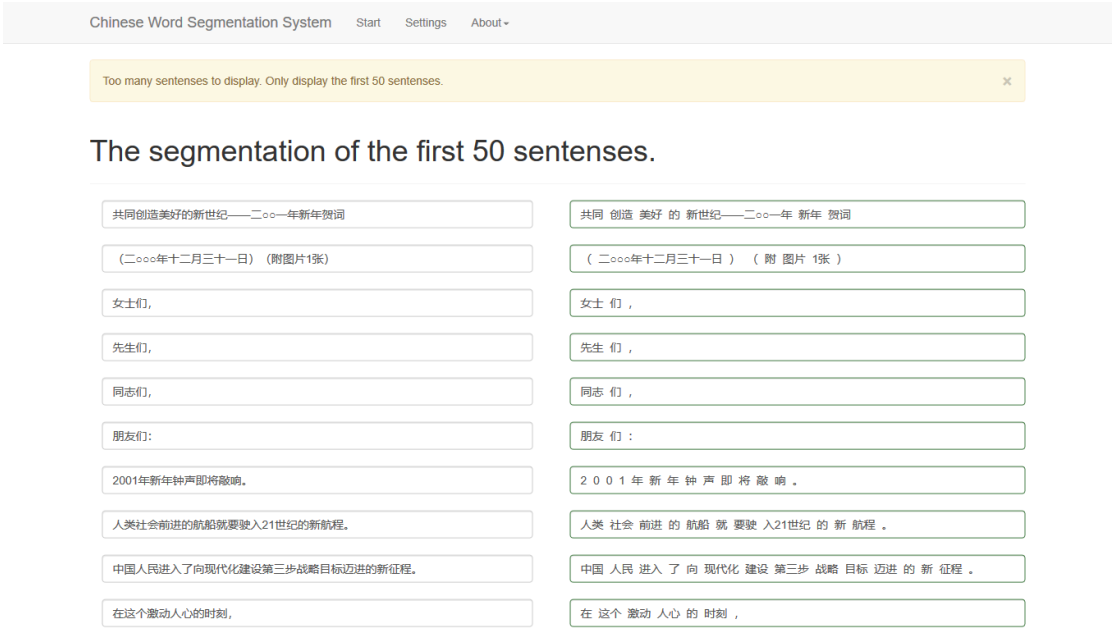


Figure 5

If your file is not encoded in UTF-8, the program won't work. (See Figure 6, Figure 7.)

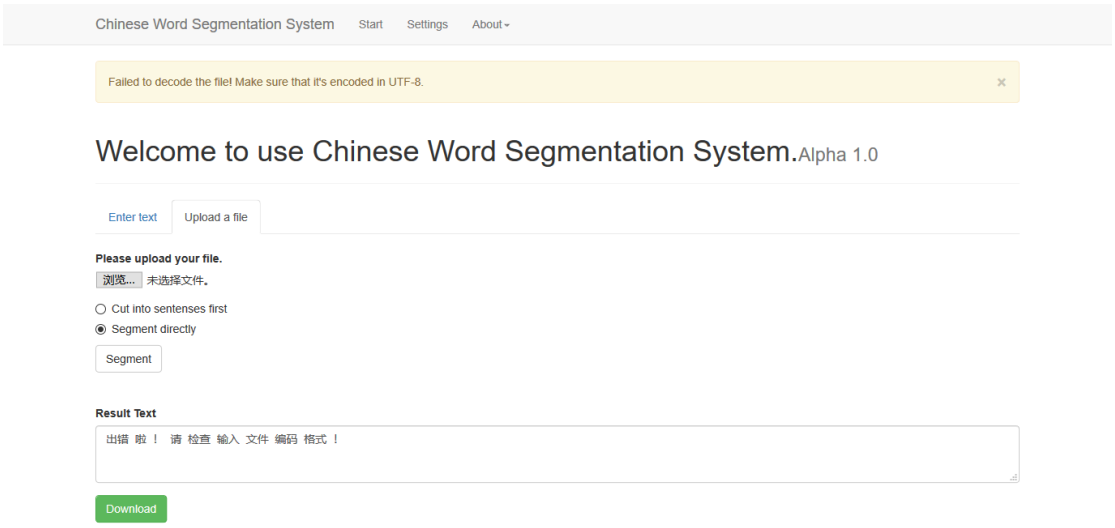


Figure 6

Figure 7

Users can modify the settings to correct the segmentations. (See Figure 8, Figure 9.)

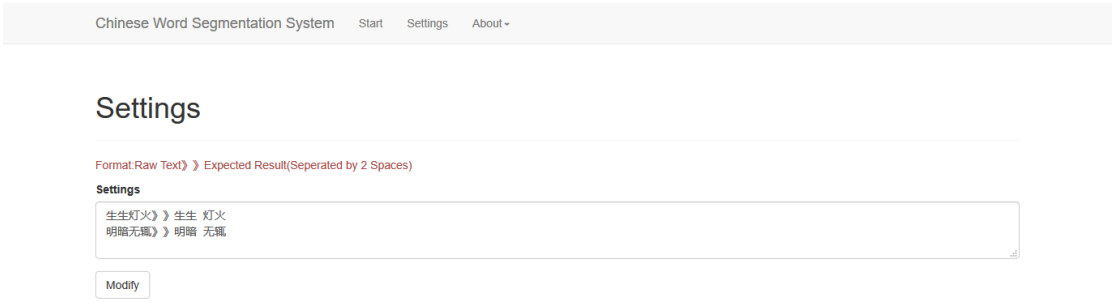


Figure 8

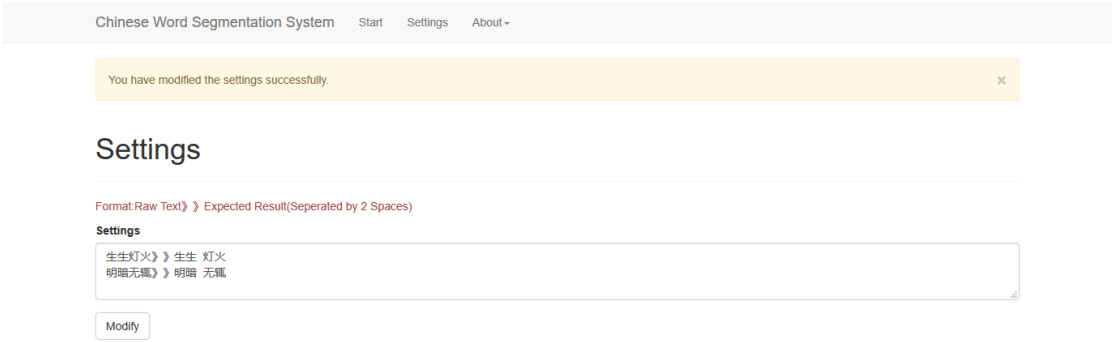


Figure 9

But if the modified settings are invalid, for example, in a wrong format, they will be ignored. (See Figure 10, Figure 11.)

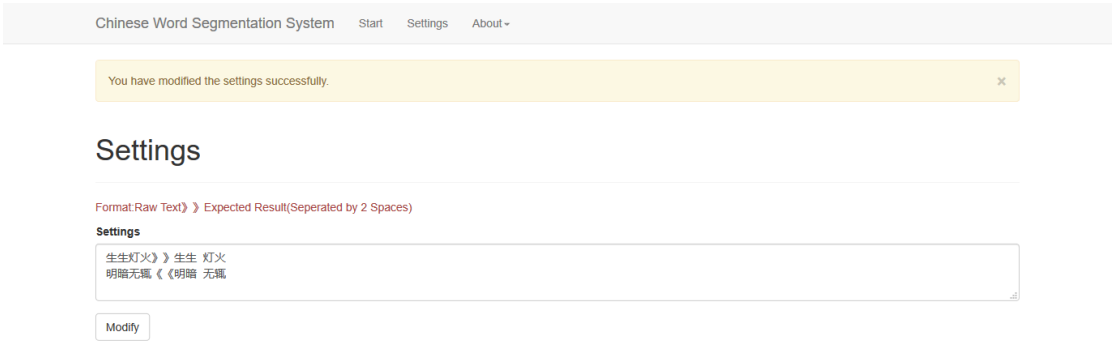


Figure 10

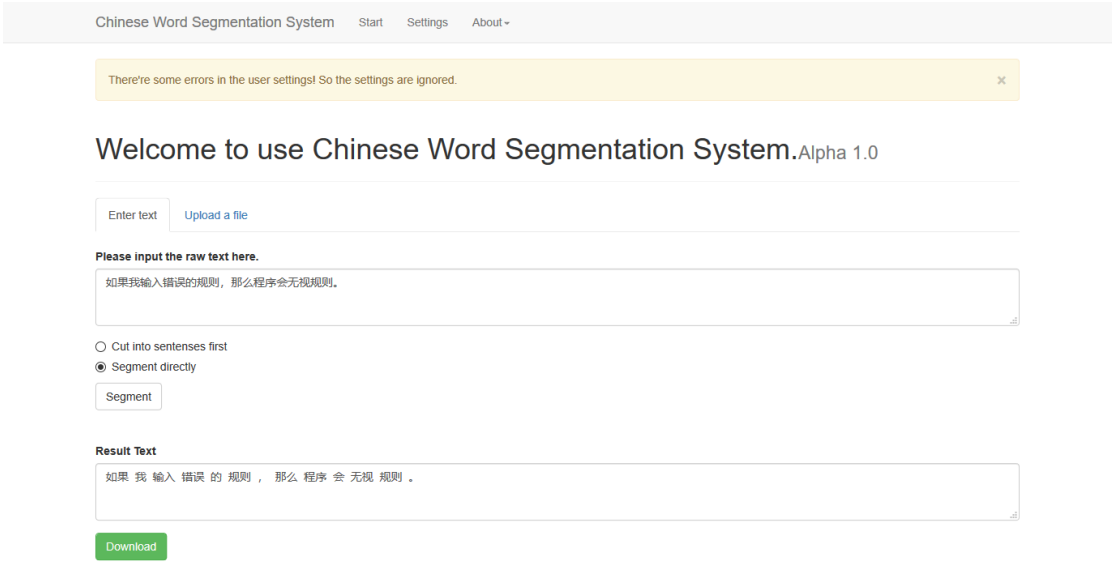


Figure 11

And the pages of Help and Copyright look like these. (See Figure 12, Figure 13.)

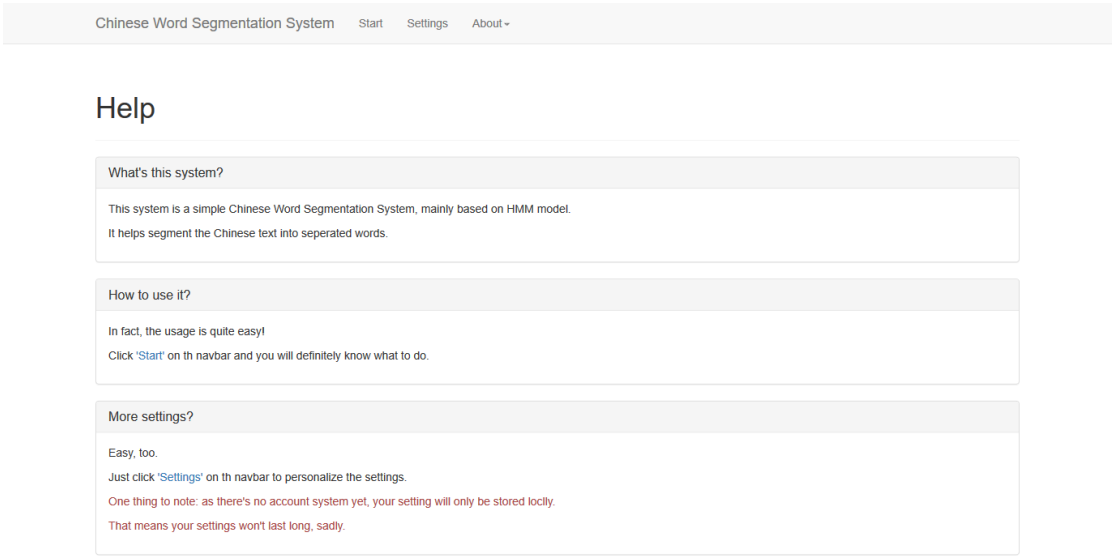


Figure 12

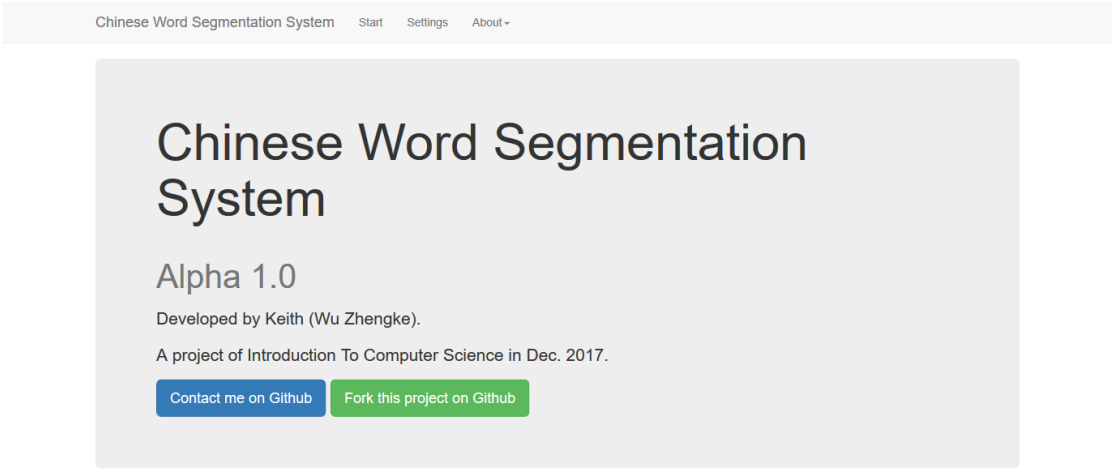


Figure 13

5.2 Test Result:

Using the HMM model trained with given training data (msr_training.utf8 and pku_training.utf8), the program segments the two testing data. The F1 Scores of this program, this program (only HMM) and Jieba (a famous Python Chinese word segmentation module) are listed as follows:

	msr_test.utf8	pku_test.utf8
This program	0.889(Figure 14)	0.829(Figure 15)
This program (only HMM)	0.793(Figure 16)	0.763(Figure 17)
Jieba	0.815(Figure 18)	0.818(Figure 19)

And for detailed results, the following figures show the complete test information.

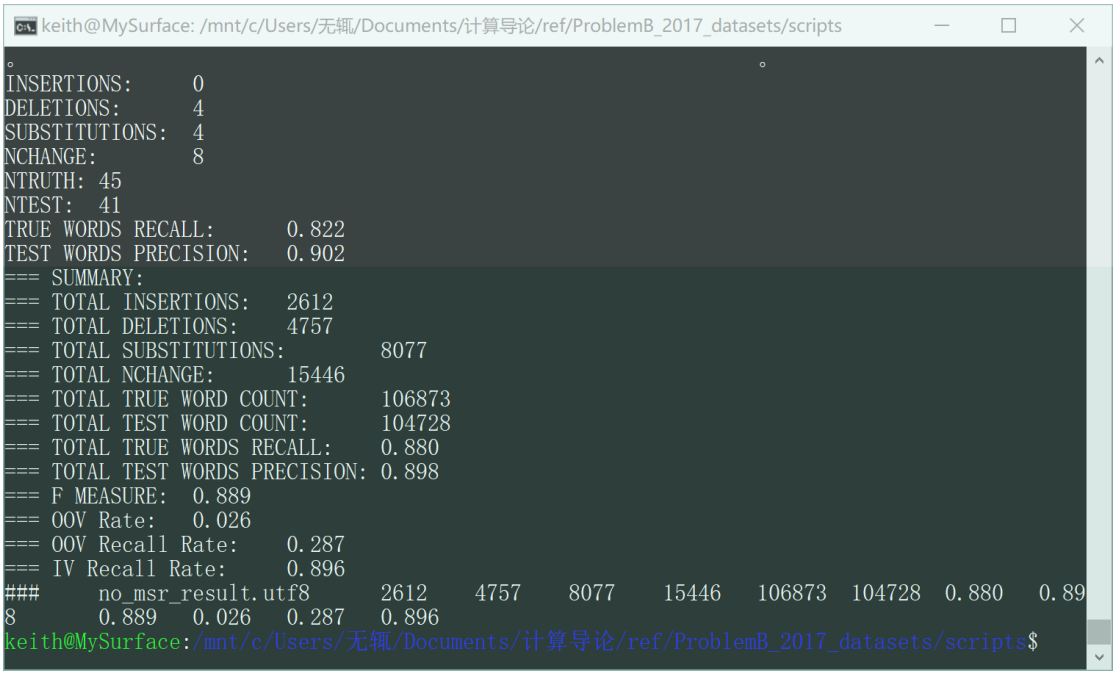


Figure 14

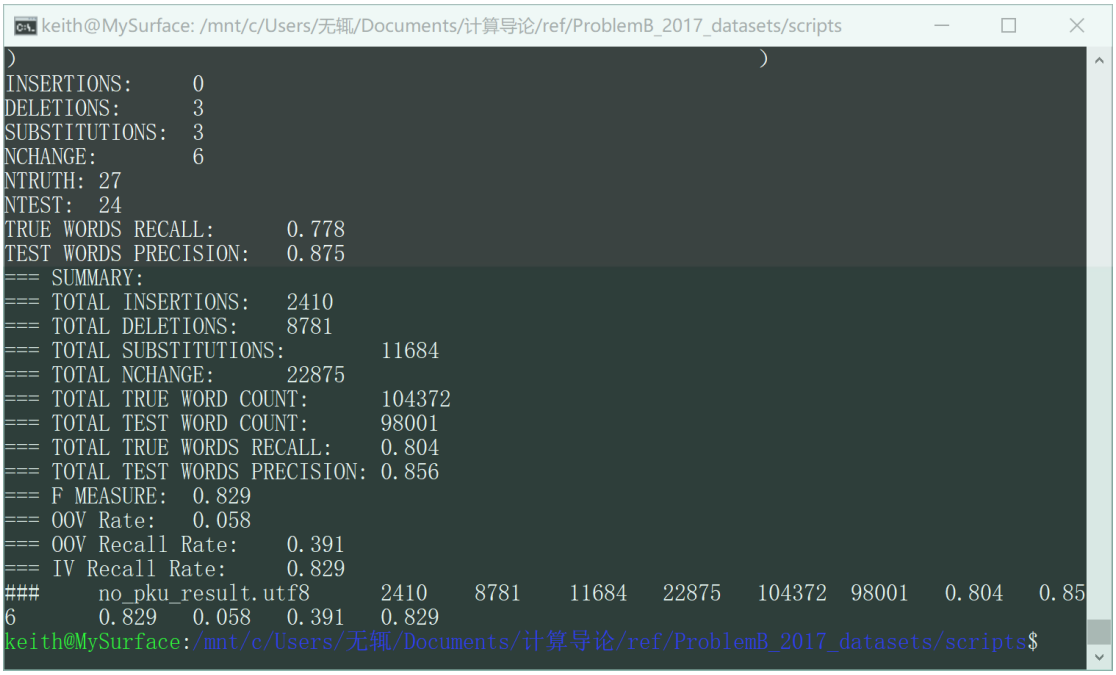


Figure 15

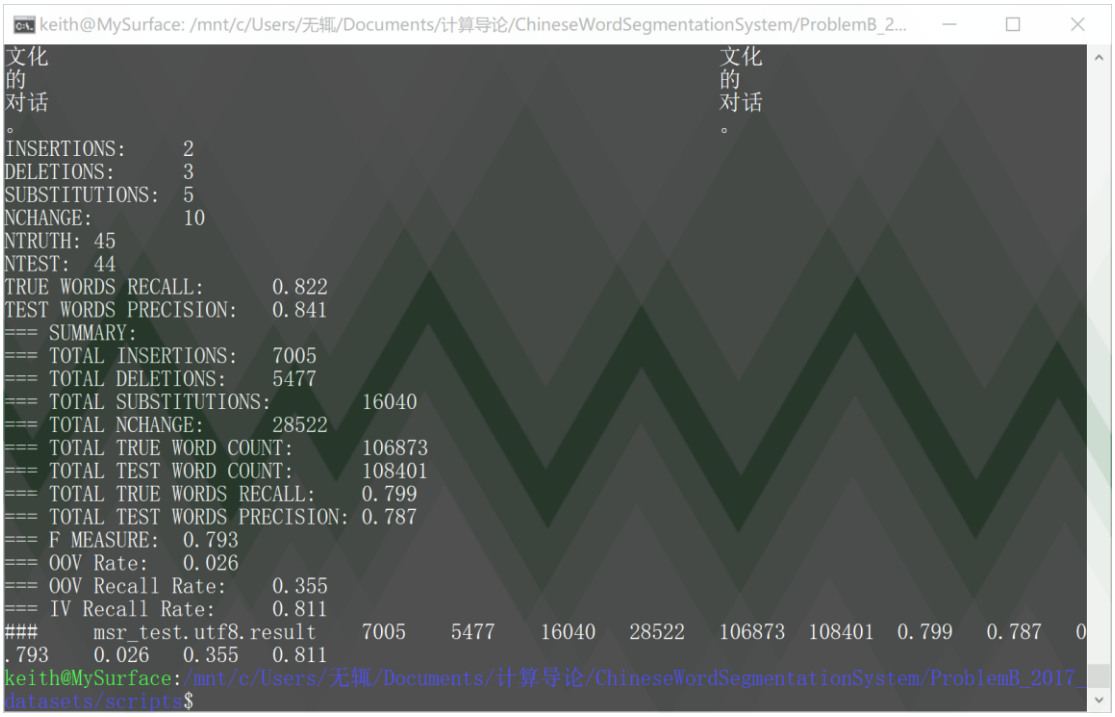


Figure 16

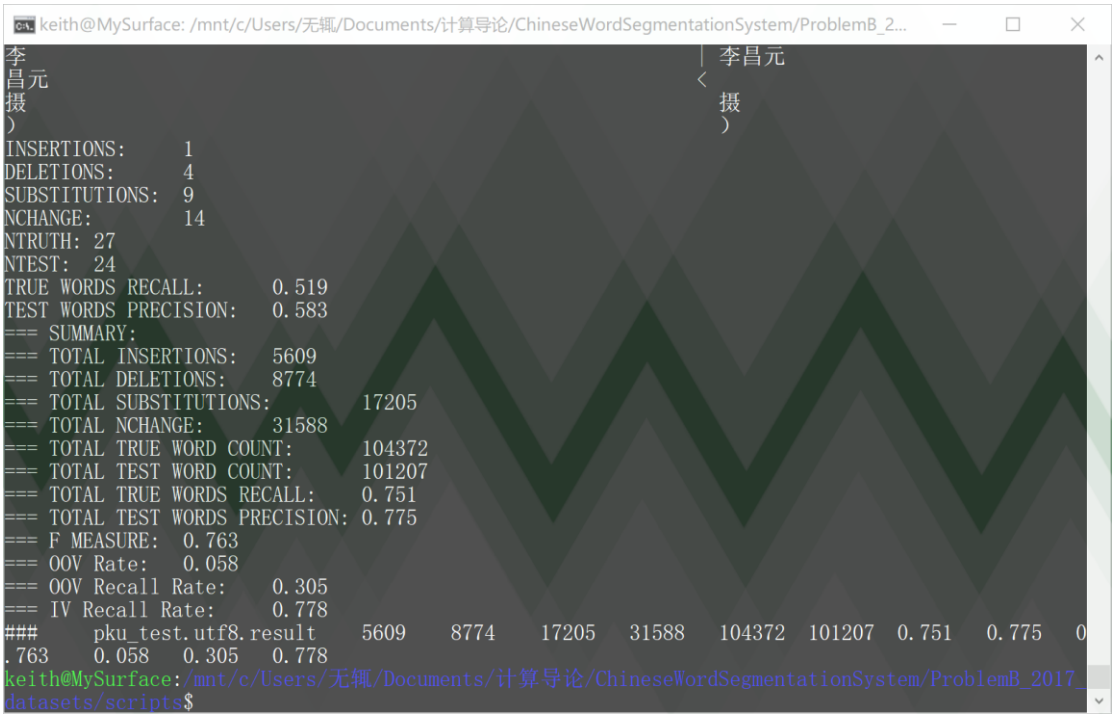


Figure 17

Figure 18

Figure 19

6. Conclusion

In comparison with the F1 Score of Jieba, it's evident that this program with only HMM is not good enough in terms of the accuracy of segmentation, not to mention other effective algorithms. To solve this problem, I thought and tried a lot. In the process, I have a much deeper insight into different algorithms about NLP and Chinese Word Segmentation.

In fact, when doing this project, I was 'forced' to learn plenty of new things, and that's exactly the meaning of this project. For example, when dealing with the segmentation, I spent lots of time learning the HMM and Viterbi Algorithm.

And as for the part of User Interface, I learned a completely new skill: Python Web Development with Flask. To implement a relatively beautiful website, I also learned the basic knowledge of HTML, CSS and the web frame Twitter Bootstrap, which took me more than a week's time in total.

In conclusion, doing this project improves my coding ability greatly and provides me with opportunities to learn new things.

7. Reference

When doing this project, I refer to quite a few articles on the Internet and some books. Some of them are listed as follows:

- 中文分词的 python 实现-基于 HMM 算法 - CSDN 博客
- 中文分词之 HMM 模型详解 - CSDN 博客
- Flask Web Development: Developing Web Applications with Python (Miguel Grinberg)
- Flask Documentation (0.12)
- Bootstrap 教程 | 菜鸟教程
- 中文分词入门之最大匹配法 | 我爱自然语言处理

With my sincere gratitude!