Title: Chinese Idioms (Final Project)

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Final Project: Chinese Idioms

Code Submitted By: Erik Andersen

Team Member Contributions:

Jin: add feature.py, animal features, sentiment features, pinyin features

I took part in brain storming the idea of the project, mainly worked on the add_feature.py, in which I added the animal zodiac information, character number, sentiment information in the corpus. I segmented the pinyin, I used the annotated dataset of Dalian University of Technology to train Naive Bayes classifier to classify the sentiment of the idioms that don't have the sentiment information. We supported the user to search for idioms that are about zodiac animals. Chinese idioms usually have multiple word to refer to the same kind of animal. For example, the "彘/pig" in "杀彘取子", the "豕/pig" in "豕突狼奔", and the "猪/pig" in "牧猪奴戏" are all referring to pig. In our search system, all those idioms can be retrieved for the user by only checking "pig" in the zodiac animal box. However, the characters that refer to animals also appear in people or place names, such as "司马昭", which contains "马/horse", we don't want to retrieve this kind of idioms but at this time, I wasn't able to find an effective NER tool to recognize proper names in idioms.

Kun: add_feature.py, difficulty features

I worked on building features of the project. In order to figure out the difficulty of each idiom, I need to find a benchmark to compare. There are many methods to determine the difficulty of idioms. For instance, we can tell an idiom is hard based on complexity of the character, or we can find frequency using of the idioms.

Then I find a very useful document HSK. HSK has collected news texts from World Forum website. The corpus contains a total of 27965 news articles with more than 20 million Chinese characters. Those news articles fall into more than 15 categories which includes topics such science, culture, technology. Based on those news articles, HSK has identified the four characters in the news corpus. They also found the frequency of each idiom and posted the result to Chinese text computing website. I use the results as a benchmark to divide our corpus into "Easy", "Medium", and "Hard". To be more specific, I first mapped all the HSK corpus with our corpus and if any idiom which is not showed in HSK corpus, I assigned value 1 to its frequency. After mapping, I iterated our corpse using frequency filed and divided idioms into three parts.

Xiaojing: index.py, query.py, templates

I worked on building the index of elasticsearch and using conjunctive search to query the fields in the corpus. I also worked on html templates to display the query results. When I tried to build the index for elasticsearch, I initially met several issues on how to properly index and query the corpus. One main issue that took me quite a long time was forgetting to import the index in the query.py, and this led to the no connection in the elasticsearch. Because of the unclear error message, I did spend a long time to finally figure out the problem.

I also took part in displaying query results on the html pages. When I tried to display our result on localhost, I integrated Bootstrap in the main page (query_page). With the help of Bootstrap, the html page could be formatted better. But due to the complex interaction, I didn't have enough time to format the other pages with Bootstrap as well.

Erik: chengyu_index.py, translation features (including modifications on template files)

One of the biggest challenges for me was integrating translation into the project. This feature was crucial in order to make the project accessible to English-speaking audiences. For indexing, the difficulties mostly lay first in finding corpora that would be easily processed (text files). This is why I used cedict 1 0 ts utf-8 mdbg.txt, as it was a dictionary with over 100,000 entries that could easily be turned into a json file. Another difficulty lay in preparing and indexing all the translation aspects to be ready for the elasticsearch index, as the chengyu text file and Chinese-English text file dictionary were formatted slightly differently. After using the Stanford CoreNLP system to segment the Chinese sentences as best we could, I stored the regular features and translation-based keys separately in order to save some space in the json files. On the query end, it was very difficult both to properly display the hover box that contained possible translations, which required a lot of editing for each HTML template file. Furthermore, it was difficult to link the translations with their Chinese counterpart, i.e. highlight the corresponding Chinese word when an English search is performed. However, I managed to employ a system that made use of this feature, which will give English speakers more access to our project. Given more time, I would like to add more to the translation features, perhaps even creating a more robust and accurate translation system. There are also some small formatting issues I found with some individual article pages that I would like to examine more closely, and ultimately fix (see 显而易见), as I was unable to find out why these individual pages were displayed as such.

Description: For this project, our goal was to create an effective information retrieval system for Chinese idiomatic expressions. Chinese idioms are incredibly diverse and numerous, and the term *idiomatic expression* has many possible translations in Chinese, including 成语 [chéngyǔ] (a set phrase usually containing 4 or 5 characters) and 俗语 [súyǔ] (a vernacular idiom). Chinese also makes use of other types of phrases, such as 歇后语 [xiēhòuyǔ] (two-part allegorical sayings). Our starting index contains 13,279 such expressions, which contains name, pinyin (with diacritical marks flattened), and description as structured data. For unstructured data, we used a regular expression to grab important elements found in the description, such as usage and source. To Chinese people, selecting a correct idiom to use is very important, so our search mechanism contains features to filter based on aspects important to Chinese culture

(such as a specific animal, or positive/negative connotation). As an added bonus, we also included translation features to aid Chinese learners.

Dependencies:

Python 3.6 Flask 1.0.2 Elasticsearch 6.7 Elasticsearch-dsl zhon==0.2.0 thulac==0.2.0 nltk==3.3 pandas==0.24.1

Build Instructions: Before running, the module flask needs to be installed. The process can be done as follows.

1. pip3 install Flask

Some sources also suggest installing the virtual environment, which can be done as follows.

1. pip3 install virtualenv

Also, please make sure that nltk is installed. nltk can be installed as follows. sudo pip3 install -U nltkthe

Next, the CoreNLP and ElasticSearch servers both need to be started. First download coreNLP version 3.9.2 from the Stanford website: https://stanfordnlp.github.io/CoreNLP/

Then, run the coreNLP server with the following command (make sure to be in the directory corresponding to the coreNLP package you downloaded): java -mx3g -cp "*" edu.stanford.nlp.pipeline.StanfordCoreNLP -props

StanfordCoreNLP-chinese.properties -file chinese.txt -outputFormat text

java -Xmx4g -cp "*" edu.stanford.nlp.pipeline.StanfordCoreNLPServer \

- > -serverProperties StanfordCoreNLP-chinese.properties \
- > -preload tokenize,ssplit,pos,lemma,ner,parse \
- > -status_port 9001 -port 9001 -timeout 15000

Then, run the elasticsearch server, making sure to be in the directory where you downloaded elasticsearch to:

./bin/elasticsearch

Files to be Submitted: (excluding this file)

Python Files: chengyu_index.py, add_feature.py, index.py, query.py HTML Files: query_page.html, page_SERP.html, page_targetArticle.html

Txt Files: 13279_chengyu.txt, cedict_1_0_ts_utf-8_mdbg.txt, HSK.txt

Xlsx Files: sentiment_vocab.xlsx

Json files: test.json

Run Instructions:

Build the corpus index:

python chengyu_index.py

Add features:

python add features.py

Build the elasticsearch index

python index.py

Run the elasticsearch query

python query.py

Index Build Time: (if applicable) - [approx. 3 minutes for chengyu_index.py, 18 seconds for add_feature.py, around 7-10 seconds for index.py]

Modules: chengyu_build_index.py, add_features.py, index.py, query.py

chengyu_build_index.py: Builds the shelf files according to what we needed add_features.py: Processes each Chengyu in the corpus and adds the zodiac animal information, the sentiment information, the difficulty level of the Chengyu and segments the pinyin.

index.py: Passes the necessary information to the elasticsearch index

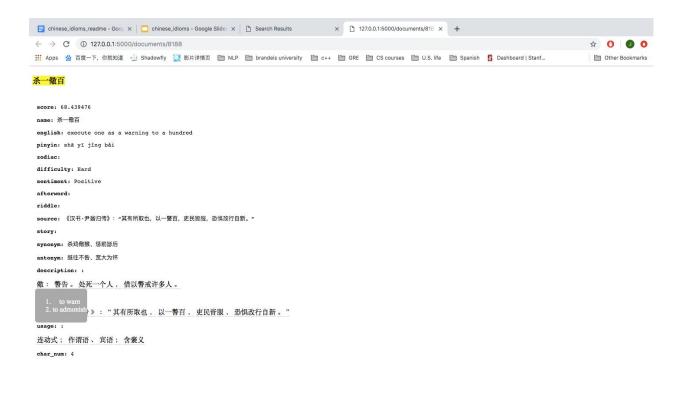
query.py: Handles the elasticsearch queries

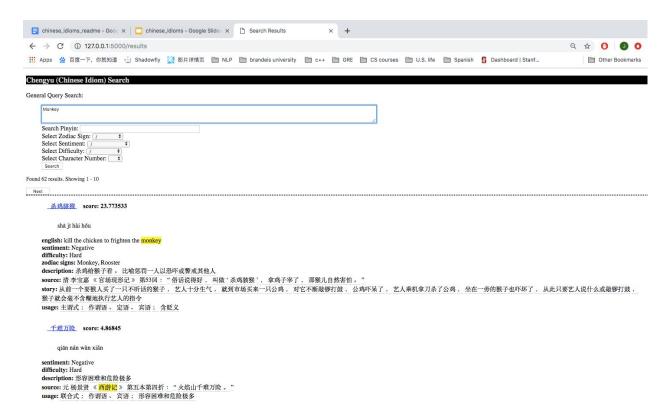
Testing:

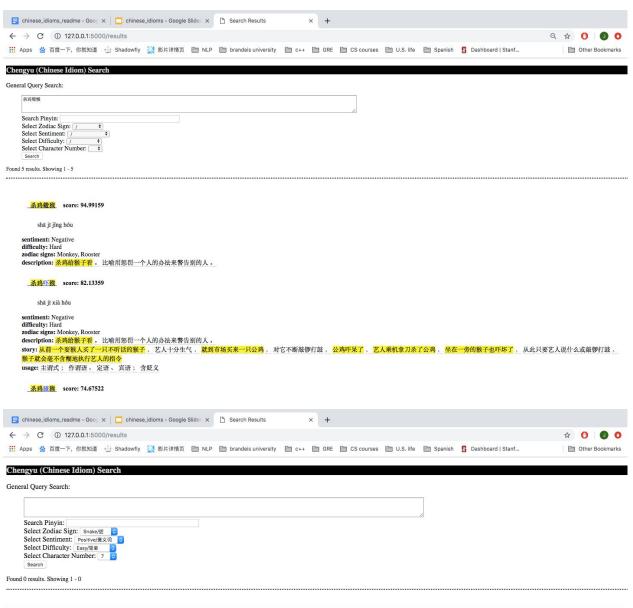
We handpicked 20 documents from our corpus, the list below contains the names of the idioms.

test_list = ["杀彘教子", "六畜兴旺", "韬光养晦","踏破铁鞋无觅处,得来全不费工夫", "恶性循环", "以其人之道,还治其人之身","鼠窜狼奔", "犬兔俱毙", "爱礼存羊", "替罪羊", "鸡烂嘴巴硬","狂吠狴犴", "阿猫阿狗", "知小谋大", "徒读父书", "前言不搭后语", "有其父必有其子", "疑人勿用,用人勿疑", "司马昭之心,路人皆知", "公说公有理,婆说婆有理"]

Test examples:







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