



北京邮电大学

Beijing University of Posts and Telecommunications

Information Retrieval & extraction course assignments

Design of search engine for animation otaku

(machine translation English
version)

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1、Project introduction

In this course design, considering the basic requirements and practicability, we decided to make a course design work of otaku oriented search engine as an information retrieval and extraction system.

We quoted the famous secondary information website - mengniang encyclopedia (<https://zh.moegirl.org/Mainpage>) The experiment was carried out on the database.



图 1 Meng Niang Encyclopedia

In this experiment, we searched the content of mengniang encyclopedia website and used saber and other commonly used animation

Roles are searched as keywords, and the return results with practical value are obtained.

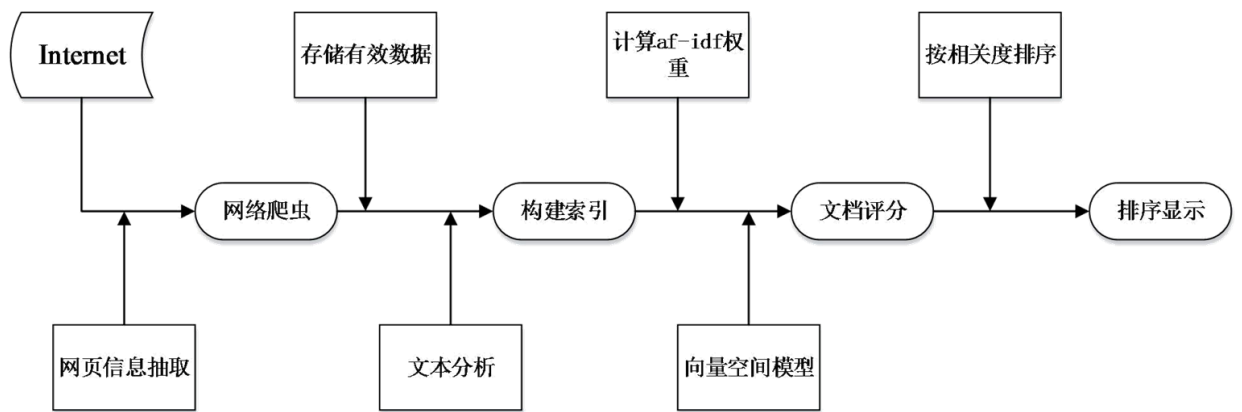
Next, we will explain in detail how we make the front-end interface, highlight it, and sort the index results in combination with our code.

We also made a simple comparison and analysis with the built-in search function on the page, and obtained some areas that need to be further improved.

2、Overall architecture

Structure analysis: we divide the task into four parts: crawling of vocabulary interpretation entry data of mengniang encyclopedia, construction of inverted index, implementation of vector space model and front-end interface of information extraction.

It is mainly divided into four modules: web crawler, index building, document scoring and sorting display. There are some sub modules between modules, including web page information extraction, data storage, text analysis, TF IDF weight calculation, vector space model modeling, correlation ranking and so on. The following is the design structure diagram caused by the whole search:

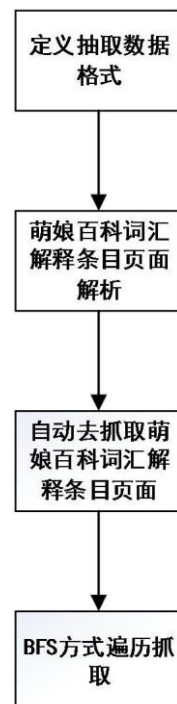


3、 Web crawler

In this job, the web crawler part uses the beautiful soup open source framework.

Beautiful soup is an HTML / XML parser written in Python. It can well handle non-standard tags and generate parsing trees. It is usually used to analyze web documents captured by crawlers. For irregular HTML documents, there are also many completion functions, saving developers' time and energy.

Flow chart of this module:



Specific implementation process of this module:

First, the source URL (https://zh.moegirl.org/Fate/stay_night) Load it into the URL manager, and the cycle starts. The judgment condition is whether there is a new URL in the URL manager. When there is a new URL, start crawling; Otherwise, end the crawler. After the crawler starts, it is necessary to use the HTML downloader to download all page contents in the source URL, and then use the HTML parser Htmlparser to parse the downloaded contents to obtain all valid URLs, body contents and titles in the body, and then load these newly obtained URLs into the URL manager again to start a new cycle. Finally, the obtained data is output by the

outputter and saved as a JSON file. At the same time, we are in the process of crawling

An exception handler is also added. If an exception occurs, a "crawl failed" message will be sent to the console to continue the next cycle.

The main codes are as follows:

```
from test import url_manager, html_downloader, html_parser,html_outputter

class SpiderMain(object):
    def __init__(self):
        self.urls=url_manager.UrlManager()
        self.downloader=html_downloader.HtmlDownloader()
        self.parser=html_parser.Htmlparser()
        self.outputer = html_outputter.HtmlOutputter()

    def crawl(self,root_url):#调度程序
        count = 1 #用count记录当前爬取的是第几个Url
        self.urls.add_new_url(root_url)
        while self.urls.has_new_url():
            #百度百科有的网页已经无法访问，需要加入异常处理
            try:
                new_url = self.urls.get_new_url()
                print 'crawl %d : %s' %(count,new_url)
                html_cont = self.downloader.download(new_url)
                new_urls,new_data = self.parser.parse(new_url,html_cont)

                self.urls.add_new_urls(new_urls)
                self.outputer.collect_data(new_data)
                self.outputer.output_html(count)#每次抓到了就搞一次
                if count == 10000:
                    break

            except:
                print 'crawl failed'
```

3.1 define the format of extracted data

For the design of crawler, we first need to determine the format of extracted data. Through demand analysis, we decide to extract the following three kinds of data. Details are as follows:

Artistic: the text of news, which is used for text analysis and index construction;

Title: the title of the news;

URL: the link of news, which is used for web display;

3.2 HTML parser

After determining the data format, next, grab a page of mengniang encyclopedia vocabulary interpretation entry, and then parse the data corresponding to the above data format. We extract page information by introducing an HTML parsing package Htmlparser in beautiful soup.

Specific implementation process: 1. Use HTML parsing package Htmlparser to parse the HTML inverted model; 2. Use the regular expression `"/% +"` to filter out the valid URL of the entry; 3. Use the find function in beautiful soup to extract the text and title in a specific `<div>`. The main codes are as follows:

```
def _get_new_data(self, page_url, soup):
    res_data = {}
    #把url也放到最终的数据中

    #<div class="lemma-summary" label-module="lemmaSummary">

    #summary_node = soup.find('div',class_="lemma-summary")
    summary_node = soup.find('div',id="mw-content-text",class_="mw-content-ltr")
    summary_node = soup.find_all('p')
    res_data['Artical']=""
    for summary in summary_node:
        ss=summary.get_text()
        res_data['Artical'] = res_data['Artical']+ss
```

3.3 HTML Downloader

HTML downloader mainly uses urllib2 module to download documents. The specific implementation code is shown in the figure below:

```
#!/usr/bin/env python
#-*- coding: UTF-8 -*-
#=====
#'''
# Created on 2016年7月8日
#
# @author: gaochuayang
#'''
#=====
import urllib2
class HtmlDownloader(object):

    def download(self,url):
        if url is None:
            return None
        response = urllib2.urlopen(url)
        if response.getcode() != 200:
            return None
        return response.read()
```

3.4 URL Manager

The URL manager can store all crawled URL addresses without repeated crawling. The specific implementation code is shown in the figure below:

Installation and deployment method: its installation and deployment is very simple. Deploy `ikanalyzer2012.jar` in the `Lib` directory of the project; `Ikanalyzer.cfg.xml` and `stopword.dic` files can be placed in the class root directory (for web projects, it is usually the `WEB-INF / classes` directory, which is the same as `hibernate`, `log4j` and other configuration files).

Main process: traverse all the TXT documents of the news subject extracted in the previous step, and save the word segmentation results in the local.

4.2 inverted index construction process

In this topic, we crawled 4287 documents, and the amount of data is not too large. Therefore, the memory based index construction method is adopted to put all the index construction process into memory for statistics, single scan, single statistical results and build the index.

Main process: traverse all 4287 documents of word segmentation in the previous step. In the loop, we count the position and number of occurrences of each word item in the document and the total number of occurrences in 4287 documents. In the specific implementation, we use two hashmaps to save the results. The HashMap uses word items as keys, and the document name, document location and other attributes are values (separated by special symbols in the middle). One is used to count the situation of each document and the other is used to count the situation of all documents.

The key codes are as follows:

```
private static void CreateIndex() {
    Set<String> stopSet = new HashSet<String>();
    try {
        stopSet = GetStopWordSet("data/stopwords.txt");
    } catch (IOException e) {
        // TODO Auto-generated catch block
        e.printStackTrace();
    }
    HashMap<String, String> hashResult = new HashMap<String, String>();
    String fileName = "";
    // 读取文件
    try {
        for (int fileIndex = 1; fileIndex <= 100000; fileIndex++) {
            fileName = fileIndex + ".txt";
            File mFile = new File("data/words/" + fileName);
            if (!mFile.exists()) {
                System.out.println(fileName + "不存在");
                continue;
            }
            HashMap<String, Integer> hashMap = new HashMap<String, Integer>();
            String content = ReadAndWrite.readFileByChars("data/words/"
                + fileName, "utf-8");
            String[] wordArray = content.split("\\|"); // "|"是转义字符 so need to add \\
            /*split 方法:将一个字符串分割为子字符串,然后将结果作为字符串数组返回*/
            for (int i = 0; i < wordArray.length; i++) {
                String str = wordArray[i].trim();
                if (str.length() > 0) {
                    String tmp = fileName + "#" + hashMap.get(str); // 最后一个为出现次数
                    if (hashResult.keySet().contains(str)) { // 包含该词
                        String value = (String) hashResult.get(str);
                        value += ("@" + tmp); // 如果包含该词说明不同的文档里都有它所以加上@来区分不同文档的value
                        hashResult.put(str, value);
                    } else {
                        hashResult.put(str, tmp);
                    }
                    System.out.println(fileName);
                } // 至此计算出了所有文档中所有分词出现的次数,并弄出了每个词在不同文档中出现的次数和在同一文档中出现的次数
            } catch (Exception e) {
                // TODO Auto-generated catch block
                e.printStackTrace();
            } finally { // finally它只能用在try/catch语句中,并且附带着一个语句块,表示这段语句最终总是被执行。
                if (hashResult.size() > 0) {
                    String value = "";
                    for (String str : hashResult.keySet()) {
                        String tmp = str + "&" + hashResult.get(str); // liang ge
                        // kong ge
                        String[] timeWord = tmp.split("@"); // "|"是转义字符
                        // 统计在文档中出现的总次数
                        tmp += "&" + timeWord.length; // 最后一个timeWord.length为这个词出现在了几个文档中
                        System.out.println(tmp); // word&filename#times@filename1#times &length
                        // 用@分开以后就统计出了一个词出现在了几个文档中
                    }
                }
            }
        }
    } catch (Exception e) {
        e.printStackTrace();
    }
}
```

4.3 calculate TD IDF weight

The most famous weight calculation method in the field of information retrieval, TF IDF weight calculation formula:

$$w_{t,d} = (1 + \log tf_{t,d}) \times \log \frac{N}{df_t}$$

Where DFT is the number of documents with word item T, which is called inverse

document frequency, TFT, D Is the number of times t occurs in D,

Is a quantity related to the document, which is called word item frequency. It increases with the increase of word frequency and the increase of word rarity.

```
public static String getTfidfString(String content) {
    String wordKey; //要索引的单词
    int numbersOfDocs; //出现次数 这个值指的是一个词出现在了多个文档里
    String tfidfStrs; //tfidf的值
    tfidfStrs="";
    String[] arrayStrings = content.split("&");
    wordKey = arrayStrings[0];
    numbersOfDocs = Integer.parseInt(arrayStrings[2]); //出现次数
    //这个值指的是一个词出现在了多个文档里

    String[] docsListArray=arrayStrings[1].split("@");//一个词在不同的文档里面出现的次数

    tfidfStrs+=wordKey;
    for (int i = 0; i < docsListArray.length; i++) { //对于这个词出现的每一个文档都要算权值
        //即所谓的T-D
        String[] valuesStrings = docsListArray[i].split("#");
        String filenameString = valuesStrings[0]; //valuesStrings0是文档的名字
        int timesOfwords = Integer.parseInt(valuesStrings[1]); //valuesStrings1是在文档中出现的次数
        // int numOfDoc = Integer.parseInt(valuesStrings[1]);

        DecimalFormat dec = new DecimalFormat("0.0000");
        double tfidf = Tf_idf( timesOfwords, numbersOfDocs);

        tfidfStrs+="&";
        tfidfStrs+=valuesStrings[0];
        tfidfStrs+="@";
        tfidfStrs+=dec.format(tfidf);
    }
}
```

4.4 index file format

Each record of the index follows the following rules: wordkey & document name

@ TFIDF value #next#& document name @ TFIDF finally forms an inverted index

file, as shown in the following figure:

```
recorders&1653.txt#1&1
发布会&79.txt#10912.txt#103813.txt#1&3
371億&3997.txt#1&1
777-300er&2224.txt#1&1
有害物&2271.txt#5&1
comaco&47.txt#10179.txt#40536.txt#40600.txt#4&4
茶点&156.txt#10186.txt#103082.txt#1&3
4万&652.txt#10767.txt#10853.txt#101285.txt#201424.txt#101674.txt#101699.txt#101911.txt#102024.txt#102174.txt#102498.txt#102
3个&200.txt#10261.txt#10481.txt#10703.txt#101349.txt#101734.txt#101738.txt#101900.txt#1&8
做饭&230.txt#102063.txt#1&2
1211年&2483.txt#203797.txt#1&2
3中&919.txt#101923.txt#102442.txt#1&3
手活&538.txt#101392.txt#101446.txt#101793.txt#102202.txt#102713.txt#102858.txt#103276.txt#1&8
4世&944.txt#101479.txt#101551.txt#101792.txt#101911.txt#102027.txt#102032.txt#202212.txt#102237.txt#402414.txt#102706.txt#10
无弹性&4178.txt#1&1
钱包&962.txt#101072.txt#101936.txt#1&3
挂在&44.txt#2045.txt#20530.txt#10700.txt#10833.txt#10858.txt#301036.txt#101133.txt#1&8
5万&1124.txt#101142.txt#101331.txt#101463.txt#101674.txt#101832.txt#101911.txt#101974.txt#102086.txt#102174.txt#102267.txt#
入社&131.txt#10603.txt#10857.txt#20911.txt#10913.txt#10964.txt#201275.txt#101399.txt#101424.txt#201455.txt#101517.txt#10169
sanyo&1142.txt#6&1
4个&46.txt#10261.txt#10347.txt#10446.txt#10502.txt#10741.txt#101521.txt#101812.txt#102144.txt#1&9
劣势&378.txt#20435.txt#10943.txt#203813.txt#1&4
1212年&2101.txt#102483.txt#102862.txt#103446.txt#2&4
5世&495.txt#10993.txt#101076.txt#201428.txt#201479.txt#101572.txt#101607.txt#101792.txt#101892.txt#101911.txt#301959.txt#10
动到&1022.txt#201856.txt#104270.txt#1&3
形成期&1450.txt#101661.txt#1&2
物种&214.txt#10250.txt#20316.txt#10362.txt#10957.txt#1&5
0人&3982.txt#104077.txt#1&2
切断&195.txt#10211.txt#10291.txt#20549.txt#10641.txt#20712.txt#101052.txt#201189.txt#101345.txt#101444.txt#101744.txt#10226
6万&944.txt#101699.txt#101911.txt#102174.txt#102266.txt#102633.txt#102753.txt#102955.txt#103065.txt#104213.txt#104244.txt#10
入神&1856.txt#1&1
5个&1.txt#101738.txt#1&2
扶桑&133.txt#10446.txt#30502.txt#30568.txt#10650.txt#101060.txt#101349.txt#101521.txt#301650.txt#201749.txt#102943.txt#1032
打沉&330.txt#1&1
十七岁&364.txt#1&1
没有关系&240.txt#10829.txt#101022.txt#101090.txt#101544.txt#1&5
6世&993.txt#101076.txt#201162.txt#101331.txt#101428.txt#101462.txt#101792.txt#201885.txt#101911.txt#201959.txt#102489.txt#10
冈田&33.txt#1053.txt#10568.txt#1&3
技术&5.txt#1021.txt#2059.txt#1064.txt#2095.txt#10102.txt#20113.txt#20230.txt#20239.txt#20253.txt#10281.txt#10297.txt#10340
xvii&3754.txt#103863.txt#1&2
1人&21.txt#10230.txt#20291.txt#10343.txt#10537.txt#20538.txt#10544.txt#10609.txt#10621.txt#10633.txt#10659.txt#30702.txt#10
6.txt#103450.txt#103529.txt#103571.txt#103604.txt#103605.txt#103639.txt#103649.txt#103709.txt#303782.txt#103806.txt#103830.
```

5、 Vector space modeling

5.1. Concept introduction

Vector space model (VSM) was proposed by Salton et al. In the 1970s, and

It is successfully applied to the famous smart text retrieval system. The processing of text content is simplified to vector operation in vector space, and it expresses semantic similarity by spatial similarity, which is intuitive and easy to understand.

The concept of VSM is simple, which simplifies the processing of text content into vector operation in vector space, and it expresses semantic similarity by spatial similarity, which is intuitive and easy to understand. When a document is represented as a vector in document space, the similarity between documents can be measured by calculating the similarity between vectors. The most commonly used similarity measure in text processing is cosine distance.

MA An unordered feature item T_i , root / word / phrase / each other document D_j can be represented by a feature item vector

$(A_{1j}, A_{2j}, \dots, A_{mj})$ weight calculation, N training documents $a_m * n = (A_{ij})$
document similarity comparison 1) cosine calculation, the advantage of cosine calculation is that it is exactly a number between 0 and 1. If the vectors are consistent, it is 1, and if orthogonal, it is 1

是 0, which conforms to the characteristic of similarity percentage. The calculation method of cosine is the product of vector inner product / module of each vector. 2)

Inner product calculation, direct calculation of inner product, low calculation intensity, but large error.

Vector space model is an algebraic model applied to information filtering, information retrieval, indexing and correlation evaluation. A document (Corpus) is regarded as a multiple meta vector space formed by index words (keywords). The set of index words is usually a phrase that appears at least once in the document. When searching, the input search term is also converted into a vector similar to a file

The model assumes that the correlation between the document and the search word can be known by comparing the angle deviation between each document (vector) and the search word (vector). A cosine of zero indicates that the search word vector is perpendicular to the document vector, that is, it does not match, that is, the document does not contain this search word.

Through the above vector space model, text data is transformed into structured data that can be processed by computer, and the similarity problem between two documents is transformed into the similarity problem between two vectors.

5.2 problem description

5.2.1 system tasks

The task of the text classification system is to automatically determine the category associated with the text according to the content of the text under a given classification system. From a mathematical point of view, text classification is a mapping process, which maps the text without category to the existing category. The mapping can be one-to-one mapping or one to many mapping. Because usually a text can be the same Multiple categories are associated and expressed by mathematical formula as follows: $F: A \rightarrow B$, where. A is the text set to be classified. B is the category set in the classification system.

The mapping rules of text classification are the discriminant formulas and rules established by the system according to the data information of several samples of each type, and then determine the text related categories when encountering new text.

5.2 evaluation method

Because text classification is fundamentally a mapping process, the marks of evaluating text classification system are the accuracy of mapping and the speed of mapping. The speed of mapping depends on the complexity of mapping rules, and the reference for evaluating the accuracy of mapping

It is the classification result of the text after expert thinking and judgment (here, it is assumed that the manual classification is completely correct and personal thinking is excluded)

The closer the results are to those of manual classification, the higher the accuracy of classification is.

```
static String filepath = "C:\\test\\index.txt";//倒排索引文件的目录位置
static HashMap<String, LinkedList<String>> WDocs = new HashMap<String, LinkedList<String>>();// <Term, Documents>
static HashMap<String, Document> docs = new HashMap<String, Document>();//文档向量
static HashMap<String, Double> queryWeight = new HashMap<String, Double>();// <Term, Weight>
```

Define filepath of string type to point to the directory location of inverted index file, define wdocs of HashMap type as some attributes in the file, and define docs of HashMap type as document vector
Define docs of HashMap type as weight value

1. load()

```
public static void Load() {
    new HashMap<String, LinkedList<Document>>();
    BufferedReader br = null;
    int n = 0;
    File f=new File(filepath);//changed by jingtiangao
    try {
        FileInputStream in=new FileInputStream(f);
        InputStreamReader r=new InputStreamReader(in, "utf-8");
        br=new BufferedReader(r);
        //br = new BufferedReader(new FileReader(filepath));

        String line;
        LinkedList<String> list;

        int z = 0;
        while ((line = br.readLine()) != null) {
            n++;
            list = new LinkedList<String>();
            // System.out.println(line);
            // System.out.println(line);
            String[] msg = line.split("&");
            String term = msg[0].trim();// TERM
            for (int i = 1; i < msg.length; i++) {
                msg[i] = msg[i].replace("#next#", "");
                String[] info = msg[i].split("@");
                String doc = info[0].trim();// DocumentTitle trim 为去掉字符左右的空格

                double weight = Double.valueOf(info[1]);//weight

                Document dos = docs.get(doc);//doc是这个文档的名字str
                if (dos != null) { // dos是这个文档的向量
```

The main function of load() is word segmentation. First read in the file.


```

for (int i = 1; i < msg.length; i++) {
    msg[i] = msg[i].replace("#next#", "");
    String[] info = msg[i].split("@");
    String doc = info[0].trim();// DocumentTitle trim 为去掉字符左右的空格

    double weight = Double.valueOf(info[1]);//weight

    Document dos = docs.get(doc);//doc是这个文档的名字str
    if (dos != null) {          // dos是这个文档的向量
        dos.setWeight(term, weight);
        docs.put(doc, dos);
    } else {
        dos = new Document();
        dos.setWeight(term, weight);
        docs.put(doc, dos);
    }
    list.add(doc);//把文档的名字加入list
}

```

In this loop body, remove the spaces around the characters.

It is defined that DOS is the document vector and DOS is the document name str. as long as there is a new document vector, set the vector weight and add the document name to the list.

```
WDocs.put(term, list);//wdoc为一个单词对应的所有的文档的表
```

Wdoc is a table of all documents corresponding to a word.

```

if = n;
System.out.println("载入内存中文档的数目为: " + n);
System.out.println("我在执行中");

```

2. query()

```
public static LinkedList<String> Query(String query) throws IOException {
    LinkedList<String> list1 = null;
    LinkedList Quer = new Segment(query).run();//对查询语句进行分词
    queryWeight.clear();
    LinkedList lis = new LinkedList();
    for (int i = 0; i < Quer.size(); i++) {
        System.out.println(Quer.get(i) + "|");
        if (WDocs.containsKey(Quer.get(i))) {

            System.out.println("倒排索引中有: " + Quer.get(i));
            if (queryWeight.containsKey(Quer.get(i))) {
                String key = (String) Quer.get(i);
                double count = queryWeight.get(key);
                count++;
                queryWeight.put(key, count);
            } else {
                String key = (String) Quer.get(i);
                queryWeight.put(key, 1.0);
            }

            lis.add((String) Quer.get(i));//lis为索引中的所有关键词
        } else {
            System.out.println("倒排索引中没有: " + Quer.get(i));
        }
    }
}
```

Firstly, word segmentation is carried out to establish a new query index in the hash table.

```
System.out.println();
KeyWords = lis;
System.out.println("索引中的关键词有: " + lis);
// Start Compute the weight of terms
java.util.Iterator<String> it = queryWeight.keySet().iterator();
while (it.hasNext()) {
    String key = it.next();
    if (WDocs.containsKey(key)) {
        int n = WDocs.get(key).size();
        double idf = Math.log10(N / n);
        double tf = queryWeight.get(key);
        double w = (1 + Math.log10(tf)) * idf;
        queryWeight.put(key, w);
        // System.out.println("Term:"+key+" weight:"+w);
    }
}
```

Computer does not have human intelligence. After reading an article, people can have a fuzzy understanding of the content of the article according to their own understanding ability; while the computer can not easily "read" the article. Fundamentally, it only knows 0 and 1, so it must convert the text into a format that can be recognized by the computer. According to the "Bayesian hypothesis", it is assumed that the words or words constituting the text are independent of each other in determining the text category. In this way, the set of words or words appearing in the text can be used to replace the text. It is self-evident that this will lose a lot of information about the content of the article, but this assumption can formalize the representation and processing of the text, and can achieve better results in text classification. Previously, in the direction of

information processing, the text is mainly represented by vector space model (VSM). The basic idea of vector space model is to represent the text by vector: (W_1, W_2, \dots, W_N) , where W_i is the weight of the i th feature item. So what can be selected as the feature item? Generally, words, words or phrases can be selected. According to the experimental results, it is generally believed that selecting words as the feature item is better than words and phrases. Therefore, in order to represent the text as a vector in the vector space, we must first segment the text by taking these words as the dimension of the vector. Represents the text. The original vector representation is completely in the form of 0, 1, that is, if the word appears in the text, the text vector is 1. Otherwise, it is 0.

Document vectorization method: expressed by TF-IDF weight.

(3) Similarity

```
public static double Similarity(LinkedList query, String doc) {
    java.util.Iterator<String> it = docs.get(doc).terms.keySet().iterator();
    LinkedList list = new LinkedList();
    double d = 0.0;
    while (it.hasNext()) {
        String term = it.next();
        d = Weight(doc, term) * Weight(doc, term) + d;
        list.add(term);
    }
    d = Math.sqrt(d); // 文档向量的模
    double q = 0.0;
    for (int i = 0; i < query.size(); i++) {
        String term = (String) query.get(i);
        double weight = queryWeight.get(term);
        q = q + weight * weight; // query的模
    }
    q = Math.sqrt(q);
    // 找出query和文档共同的部分, 做内积
    list = ANDList(query, list);
    double sim = 0.0;
    for (int i = 0; i < list.size(); i++) {
        String term = (String) list.get(i);
        sim = sim + Weight(doc, term) * queryWeight.get(term);
    }
    double score = sim / d / q;
    return score;
}
```

Calculate the similarity and add the weight of each file to the hash table.

Sorting method: sort according to the similarity calculation results according to the score. At the same time, it can be sorted according to time and time

Heat is simply sorted.

(4) LinkedList

```
public static LinkedList<String> Top10(LinkedList query, LinkedList lists) {
    Map map = new TreeMap();
    LinkedList<String> list1 = new LinkedList<String>();
    for (int i = 0; i < lists.size(); i++) {
        String docname = (String) lists.get(i);
        double score = Similarity(query, docname);
        // System.out.println("Score=" + score);
        map.put(docname, score);
    }

    List<Map.Entry<String, Double>> list = new ArrayList<Map.Entry<String, Double>>();
    map.entrySet().forEach(list::add);

    Collections.sort(list, new Comparator<Map.Entry<String, Double>>() {
        public int compare(Entry<String, Double> o1,
            Entry<String, Double> o2) {
            return o2.getValue().compareTo(o1.getValue());
        }
    });

    return new LinkedList<String>(list);
}
```

The final sorting result LinkedList is returned in the form of hash table.

6、 Search engine front end

6.1 effect display

This experiment searched the well-known animation consulting website mengniang encyclopedia. The reason why we experimented with this website is to design a tool that can help the otaku quickly obtain first-hand fresh consultation.

Here are the search results we returned:

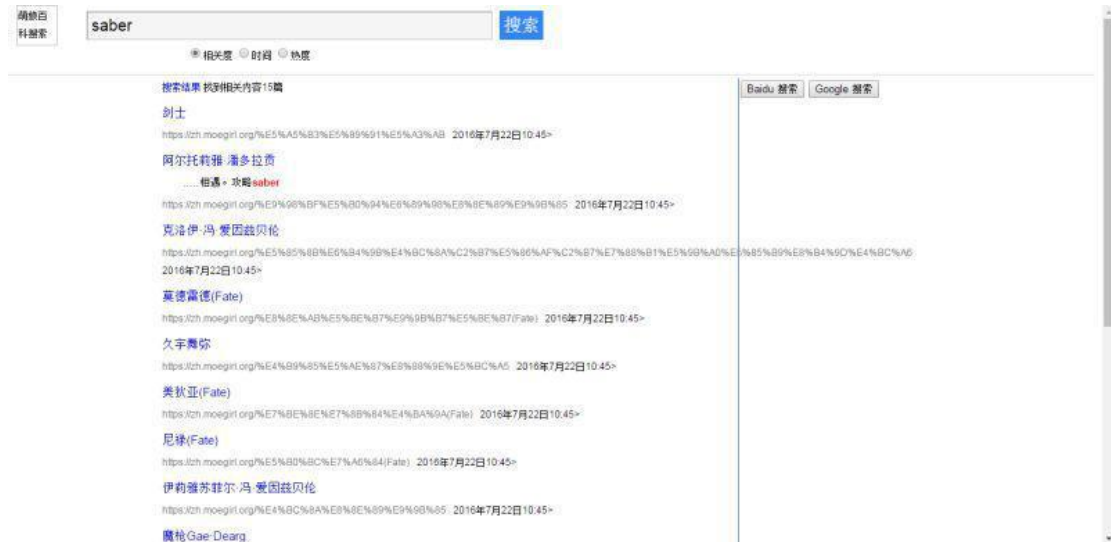


Figure 1-1 English search results

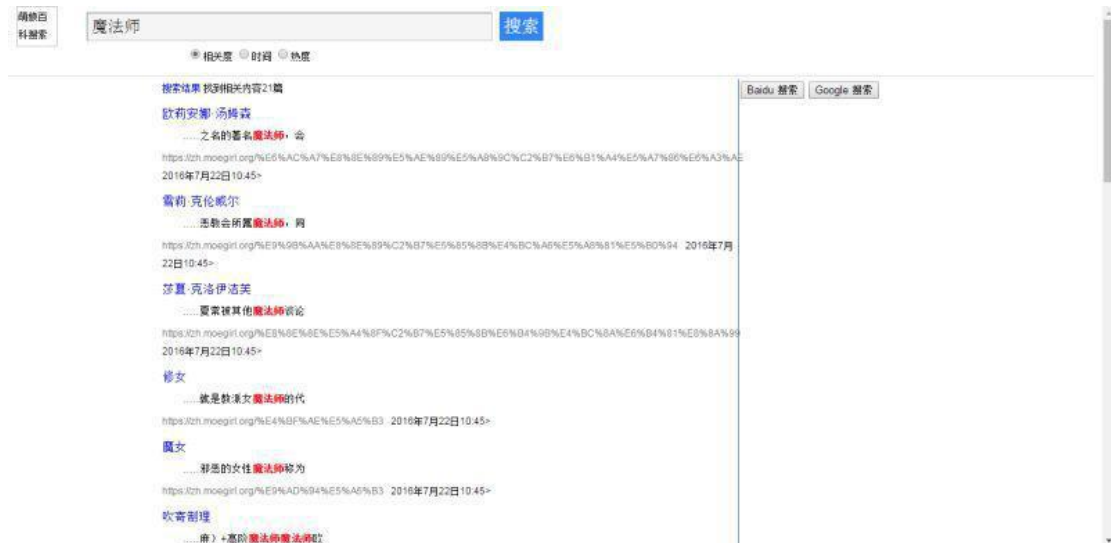


Figure 1-2 Chinese search results



Figure 1-3 multi field search results

We can also take a look at the results returned by the same keywords using the search function of the web page, as shown in the figure below. When the search content is an entry, the page will automatically enter the explanation of the entry:

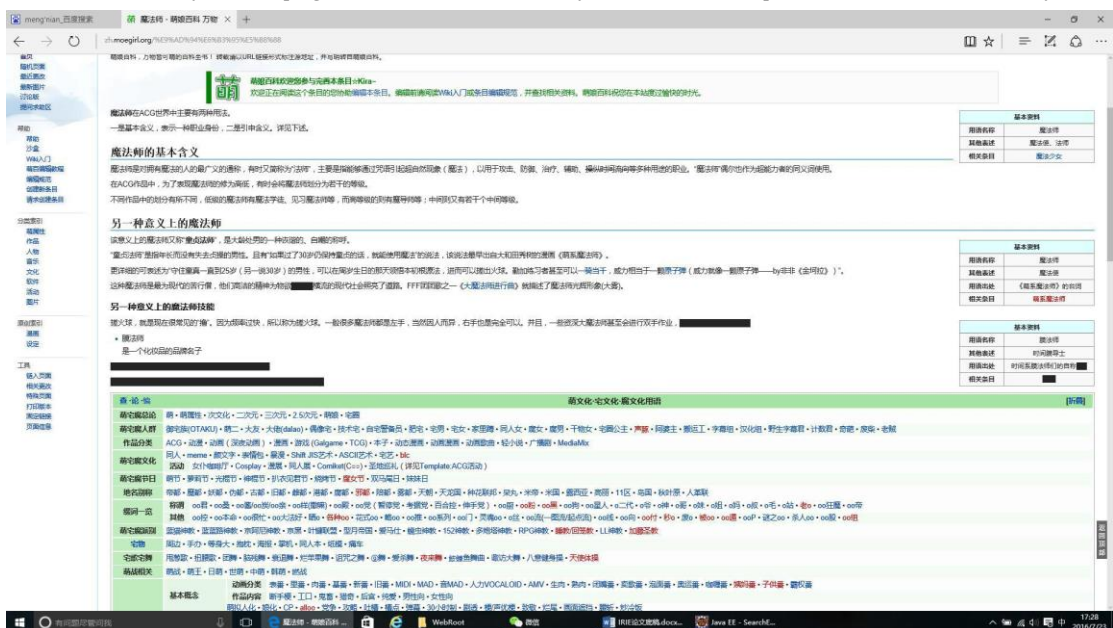


Figure 1-4 entry entry

When the input content is not a known term, the returned results are as follows:



图 1-5 search for "1984" "proposal"

Obviously, the returned results are different from our engine. We have more returned results. The official search engine

We don't know what algorithm is used in. For example, they can sort and filter search results according to the number of visits of a page, which we can't realize.

In addition, the interface of the official engine is also better than ours, which is what we need to improve. We should beautify it according to the two-dimensional enthusiasts.

6.2 explanation of front end design code segment

The following is the relevant code of our front end:

```
TreeMap<String,String> map1 = new

TreeMap<String,String>(new Comparator(){

    public int compare(Object o1, Object o2)
    { // TODO Auto-generated method stub
        return o2.toString().compareTo(o1.toString());
    }

});

//Treemap by heat, newly added
TreeMap<String,String> map2 = new

TreeMap<String,String>( new Comparator(){

    public int compare(Object o1, Object o2)
    { // TODO Auto-generated method stub
        return o2.toString().compareTo(o1.toString());
    }

});
```


We first sort the retrieved results according to the ranking obtained by VSM algorithm, from high to low heat.

When the input keyword is not empty, the absolute position of the content containing the keyword is found.

The list variable is used to record the number of returned results. When the list is null, it will prompt "nothing found"

```

}else
{
    %>

    <div class="bobyTitle">

        <div class="bobyContent">
            Nothing found!!!
        </div>
    </div>
    <%} %>

```

The design of the page is completed in HTML. Our display items include heat, number of results, etc. the implementation method is as follows:

```

<div id="head">

    < a herf = "index. JSP" id = "input_a" > < img ALT =
    "mengniang Encyclopedia Search" class = "inputimg" ></a>
    <form id="fmSearch" method="post" action="index.jsp">
        <input type="text" name="keyWord" class="inputText"
value="<%=keyword %>" />
        &nbsp;
        < input type = "submit" value = "search" class =
        "inputsubmit" / >
        <form method="post" action="index.jsp" class="formRadio">
            <div style="clear:both;margin-left:230px;">
                <input type="radio" name="radio" value="score"
Onclick = "save()" / > correlation
                <input type="radio" name="radio" value="time"
Onclick = "save()" / > time
                <input type="radio" name="radio" value="hottop"
Onclick = "save()" / > heat
            <!-- input type = "submit" value =
            "submit" / - > < / form >
        </div>
        </form>

```

According to the previous sorting, we will display the address link and profile information on the page one by one, and the corresponding code segment is shown in the figure below:

```

}

for (String o:list)
{
    //Engine.ResultModel mod = (Engine.ResultModel)o;
    //Engine.Result mod=new Result(o+"txt");
    //ucas.JsonObject myObject =
ucas.JsonUtil.readFile2JsonObject("/data/"+o+".json");

```



```

        return list1;
    }

    public String HighLightKey(String content) {
        content = content.replaceAll(" ", "");
        for (int i = 0; i < KeyWords.size(); i++) {
            String word = KeyWords.get(i).toString();
            content = content.replaceAll(word,
                "<font style='color:#ff0000;font-weight:bold;'>" + word
                + "</font>");
        }

        return content.replaceAll(
            "</font>[\\W]*<font style='color:#ff0000;font-weight:bold;'>",
            "");
    }

    public String Path(String path, String o) {

        o = o.replace(".txt", ".json");
        String data = "data\\";
    }

```

Figure 2-2 schematic diagram of search code

The implementation of information retrieval is shown in the figure above. The overall idea is that we traverse all index files on the whole website, record the pages containing the input keywords, then display the upper and lower text, highlight the keywords, and record their URLs for users to access.

Our husband becomes a link to the target URL. The page displays the inclusion sentence with keywords, in which the keywords are highlighted, as shown in Figure 2-1.

The implementation of highlighting is relatively simple. We only need to modify the color and font of the keywords recorded earlier, as shown in Figure 2-2.

For the display of the text introduced below, we use the keyword itself as the benchmark to display the characters with fixed length before and after it (the specific length will be adjusted according to the number of keywords).

In this way, we have completed the preliminary extraction of relevant information, which is actually to return the context content of keywords.

Obviously, we will return many related pages. We will display them according to the VSM sorting algorithm.

8、Problems encountered in development

8.1 problems encountered in writing crawler

When I first wrote the crawler program, I only crawled 100 Web pages as a test, so in the code, I uniformly stored the crawled data in a list, and then uniformly saved all the web pages. However, when the data is very large, it may cause heap depletion and lead to memory overflow. Therefore, later, I changed to perform persistence operation immediately after crawling a web page, and then, Clear the data in the memory in time to prevent insufficient memory space.

If a URL downloaded by the crawler is invalid, the program will crash, so exception handling is added to the circular program to prevent the crawler from stopping because of the 404 page.

The modified code is as follows:

```

def crawl(self, root_url): #调度程序
    count = 1 #用count记录当前爬取的是第几个Url
    self.urls.add_new_url(root_url)
    while self.urls.has_new_url():
        #百度百科有的网页已经无法访问，需要加入异常处理
        try:
            new_url = self.urls.get_new_url()
            print 'crawl %d : %s' %(count, new_url)
            html_cont = self.downloader.download(new_url)
            new_urls, new_data = self.parser.parse(new_url, html_cont)

            self.urls.add_new_urls(new_urls)
            self.outputer.collect_data(new_data)
            self.outputer.output_html(count) #每次抓到了就搞一次
            if count == 10000:
                break
            count = count + 1
        except:
            print 'crawl failed'
    # self.outputer.output_html()
if __name__ == "__main__":
    #root_url = "http://baike.baidu.com/view/21087.htm"

```

8.2 problems encountered in writing search engine programs

The background of our search engine is written by using java language and relying on the existing framework for secondary development, but the data structure of a page of mengniang encyclopedia obtained by the crawler is a file in JSON format redefined by ourselves, so we need to rewrite the coding and data reading and writing interfaces, otherwise null pointer exception and other exceptions will appear in the background Read in index file garbled.

In order to solve the above problems, we modified the original Unicode encoding format in the framework to UTF-8, solved the problem of garbled code, rewritten the code of the read-in part of the JSON format file, and solved the exception of data read-write null pointer. The modified code is as follows.

```

public static JSONObject readFile2JsonObject(String path) {
    String JsonContext = JsonUtil.ReadFile(path);
    JSONArray jsonArray = JSONArray.fromObject("[ "+JsonContext+" ]");
    JSONObject jsonObject = jsonArray.getJSONObject(0);
    JSONObject MyObject = new JSONObject(jsonObject.get("Artical").toString(), jsonObject.get("Time").toString(),
        jsonObject.get("Total").toString(),
        jsonObject.get("title").toString(), jsonObject.get("URL").toString());

    return MyObject;
}

```


9、 Group division of labor

Gao chuyang: development of information index and relevance ranking code, preparation of relevant documents, Yang Xue: development and testing of crawler system, preparation of relevant documents, Feng Yimeng: development and testing of information extraction system, preparation of relevant documents

Sun Chuanhao: implementation of web pages and preparation of relevant documents
The workload of the four people is equal, and the teacher can give them points as appropriate.

