# 数据挖掘实验报告

#### 1. VSM

#### 1.1 预处理

Input: 单篇文件

Processing: 去除数字, 标点, 词缀, 等, 进行分词, 生成处理过的词表

Output: 该文件生成的 wordList

#### 主要实现:

C:\Users\eric\Desktop\VSM\Vsm\preProcessing.py

```
| def DealwithFile(filename):
| #fw = codecs.open(filename, 'r', )
| f = open(filename, 'r', errors='ignore')
| text = f.read()
| #去除读取文件后的格式(换行 缩进)
| text = text.replace('\r', '').replace('\n', '').replace('\t', '')
| remove_digits = str.maketrans('', '', string.digits)
| text = text.translate(remove_digits)
| f.close()
| return text
| #(2) 表除标点和大写,数字.分词
| def remove_tokens(text):
| lowers = text.lower() # 大小写
| # 去除标点符号
| remove_punctuation_map = dict((ord(char), None) | for char in string.punctuation)
| no_punctuation = lowers.translate(remove_punctuation_map)
| tokens = nltk.word_tokenize(no_punctuation)
| return tokens

# (2) 提取词干
| def Stemming(wordlist):
| LancasterStem = nltk.LancasterStemmer()
```

### 1.2 生成词典

Input: 生成的 wordList

Processing:构造词典的数据结构

{"key1": 32}//word 和其全局的频率

Output: Dict in redis

目的: 筛选 keyword (选择频率高于 15 的词, 将词典从 80000 项目讲到 8703 项)

```
| Description | Hospital | Hospi
```

#### 1.3 TF\_IDF 计算

在统计的同时生成 TF\*IDF

#### 1.4 形成 VSM 数据

```
27.0.0.1;6379) get monet
('huang', 'o.0, 'rec.sport.hockey54081\': 5.872808483968911, \"alt.atheism64225\": 5.872808483969911, \"alt.atheism64225\": 5.872808483968911, \"comp.os.me-windows.misc10123':
11.74651697937821, \"comp.os.me-windows.misc10134\": 11.746516977937821, \"comp.os.me-windows.misc10123':
11.74651697937821, \"comp.os.me-windows.misc10144\": 11.746516977937821, \"comp.os.me-windows.misc10140\": 5.872808483969911, \"comp.os.me-windows.misc10123':
11.74651697937821, \"comp.os.me-windows.misc10140\": 5.872808483969911, \"comp.os.me-windows.misc10140\": 5.872808483969911, \"comp.os.me-windows.misc10140\": 5.872808483969911, \"comp.os.me-windows.misc10140\": 5.872808483969911, \"comp.os.me-windows.misc10123'
13.72808483969911, \"sci.apectromics54202\": 5.872808483969911, \"sci.apectromics54251\": 5.8728084839
```

## 2. KNN

#### 2.1 面向文件

先转换第一部分生成的 Dict 成为 File Version(之前工作出的纰漏,这里弥补,是因为没有想明白的问题,以后要先想明白再动手。)

#### 2.2 计算 KNN

根据已经形成的文件词典,以此计算 KNN,选取最大值,进行划分

```
for key in keys:
    if key[0:3] == '123':
        continue
    else:
        Redis.delete(key)
print('update finishing..')
files=Redis.keys()
cosList=[]
for file in files:
    cos=CalculateKNN(json.loads(Redis.get(file)),InputWordDict)
    cosList.append({file:cos})
print("KNN Processing finished")
cosList.sort()
print(cosList[0][0]," ::: ",cosList[1][0],"------ are likey to be the
```

#### 2.3 命中率

选取了 10 个文件进行随机测试,正确划分完成了 6 次,准确率 60 左右,进一步的准确率还需要更多的样本进行支撑