李桂欽 ID: R04725050 Department: 資訊管理 碩一 Homework: Name:

## **Programming Assignment 3:**

## Programming Assignment 3 (1/3)

#### Multinomial NB Classifier:

- Text collection:
  - The 1095 news documents
  - □ 13 classes (id 1~13), each class has 15 training documents.

class id training doc ids		doc id class id
1	11 19 29 113	7 2
2	1234	14 8
		22 11
13	485 520 523	23 11

- The remaining documents are for testing.
  - Generate an output file (output.bt) that records your classification
  - Exclude all training documents.
  - Ascending order to doc id.

# Programming Assignment 3 (2/3)

#### □ Note:

- For each class, you have to calculate M P(X=t|c) parameters. M is the size of your vocabulary.
- Then, the total number of parameters in your system will be |C|\*M ← can be a huge number.
- We know that many terms in the vocabulary are not indicative.
- Employ a feature selection method and use only 500 terms in your classification.
  - X2 test.
  - Likelihood ratio
  - Pointwise/expected MI.
  - Frequency-based methods
- When classify a testing document, terms not in the selected vocabulary are ignored.

# Programming Assignment 3 (3/3)

To avoid zero probabilities, calculate P(X=t|c) by using add-one smoothing.

$$P(X = t_k \mid c) = \frac{T_{ct_k} + 1}{\sum_{i' \in V} (T_{ci'} + 1)} = \frac{T_{ct_k} + 1}{\sum_{i' \in V} (T_{ci'}) + |V|}$$

- □ Please zip and submit ¹ your classification result (output.txt), 2-source code, and 3-a report to TA.
  - 3 weeks to complete, that is, 2015/12/22.
- TA will announce best micro/macro-averaging precision, recall, and F1.

## My program result:

Step1: 部署 Hw3

Step2: 在流覽器輸入 http://www.mytest.com/SearchService.php 生成的 Result 文檔, 詳見 program\_result 檔夾, 大致如下:

```
 doc_id
 class_id

 17
 2

 18
 10

 20
 2

 21
 2

 22
 2

 23
 10

 24
 10

 25
 2

 26
 10

 27
 10

 28
 2

 30
 2

 32
 2

 33
 2

 34
 10

 35
 10

 36
 2

 37
 2

 38
 10

 39
 2

 40
 10

 41
 2

 42
 10

 43
 10

 45
 2

 46
 2

 47
 2

 48
 10

 49
 10

 50
 10
```

## My program architecture:



## My program main class:

## PorterStemmer Class Structure:



## 演算法實現過程:

第一步, 處理複數, 以及 ed 和 ing 結束的單詞。

第二步,如果單詞中包含母音,並且以 y 結尾,將 y 改為 i。

第三步, 將雙尾碼的單詞映射為單尾碼。

第四步, 處理-ic-, -full, -ness 等等尾碼。

第五步,在<c>vcvc<v>情形下,去除-ant,-ence 等尾碼。

第六步,也就是最後一步,在m()>1的情況下,移除末尾的"e"。

### 演算法使用說明:

傳入的單詞必須是小寫

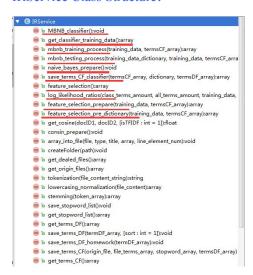
## 參考學習網站:

http://tartarus.org/~martin/PorterStemmer/

http://snowball.tartarus.org/algorithms/english/stemmer.html

http://blog.csdn.net/noobzc1/article/details/8902881

### **IRService Class Structure:**



### 類的主要函數:

```
*Multinomial Model
                                                                                                ➤ 文章分类的入口函数
  # #/
 public function MBNB_classifier() {...}
public function get_classifier_training_data(){...} 获取文章分类的training_data的information
 public function mbnb_testing process( $training_data_diffionate) training_data_diffionate, $testing_Quara, $testing_process( $training_data_diffionate, $testing_quara, $test
                                                                     ] /*...*/
 public function naive_bayes_prepare 0 (...) 在文章分类前对所有文件包括 training data 和testing data
 public function save_terms_CF_classifier($termsCF_array,$dictionary,$termsDF_array){...}
                                                       public function feature_selection(){...} 文章分类前的feature selection
 public function log likelihood_ratios( $class_terms_amount, $all_terms_amount, $training_data, $statistical_table )
                                                                                                      采用的是LLR算法
public function feature_selection_prepare($training_data, $termsCF_array) feature selection 前的准备
 public function feature_selection_pre_dictionary($training_data,$termsCF_arfonture)selection 所需要的
```

