執行環境:

ruby 1.9.3 (Windows 7)

執行方式:

在 Windows console 端執行指令 ruby pa3.rb (須安裝 ruby 1.9.3)。

• 執行時間(筆記型電腦, i5 處理器、4GB 記憶體):

	第一次	第二次	第三次
Feature Selection	8'30"	11'20"	10'00"
Training & Testing	5′30″	4'30"	4'22"
總時間	14'00"	15'50"	14'22"

平均執行時間:14'44"

• 程式執行步驟:

1. "training.txt"

首先讀取老師提供的"training.txt"檔案,把每個 class 和檔案的關係存成 {"1" => [11, 19, 29,...], "2" => [1, 2, 3,...], ...} 的 hash 形式,方便之後讀取。

```
7  # Read "training.txt" into trainingFiles{}
8
9  array = open('training.txt', 'rb').readlines
10  trainingFiles = Hash.new
11  array.each do |arr|
12  a = arr.split
13  trainingFiles[a[0]] = a.drop(1)
14  end
```

2. Extract vocabulary

把 training document 的內容讀取出來,並且用之前寫的 extract 函式(寫在"extraction.rb"檔案中)做關鍵字的萃取,刪除多餘的 stopword 以及做 stemming,把初步的關鍵詞彙存在 *vocab* 這個變數中。

```
# Extract vocabulary from training document set (require extraction.rb)

print "Extracting vocabulary from training documents..."

vocab = Array.new

trainingFileTerms = Hash.new

trainingFiles.each { |key, docNames|

docNames.each do |dn|

f = open( INPUT_DIR + '/' + dn + '.txt', 'rb').read

extracted_array = extract(f)

trainingFileTerms[dn] = extracted_array

vocab.concat(extracted_array)

end

vocab = vocab.uniq
```

3. Feature selection

呼叫 selectFeatures 函式(寫在"featureSelection.rb"檔案中)。這邊選擇使用投影片中的「 χ^2 feature selection」方法,針對每一個 class,對前面 *vocab* 變數中的每一個 term 計算出現於該 class 的次數、出現於非該 class 中的次數,再計算 χ^2 統計值,並且在每個 class 中選出前 38 或 39 個 χ^2 統計值最高的 term(比較重要、比較有判別性的 term),放入最後真正會使用的關鍵字集合中。最後會湊滿 500 個關鍵字,存在 *vocabFS* 這個變數中。

```
# Feature selection: chi-square (require featureSelection.rb)

vocabFS = Array.new

for i in 1..CLASS_NUM

res = selectFeatures(trainingFiles, trainingFileTerms, vocab, CLASS_NUM, i.to_s)

vocabFS.concat(res)

end
```

```
def selectFeatures(trainingDocs, trainingDocTerms, vocab, class_num, c)

print "\nFeature Selection for class " + c + "..."

termUtility = Hash.new

#for each term in vocabulary

vocab.each do |t|

form = Hash.new

x = 0.0

countDocs = 0.0

#for each document in class c, check if t exists

trainingDocs[c].each do |dn|

if trainingDocTerms[dn].include?(t)

x += 1

end

countDocs += 1

end

form["on&present"] = x

form["on&present"] = countDocs - x

on = countDocs

y = 0.0
```

4. Training

確定關鍵字集合之後就進入 training 階段。針對每一個 class,計算每一個 term 在該 class 中出現的條件機率,也就是說會有 C * M = 13 * 500 = 6500 個條件機率值,存在 condprob 這個變數中,以{"apple" => [0.01, 0.032,...], "tree" => [0.1, 0.032,...], ...} 這樣的 hash 形式儲存。

```
for i in 1..CLASS_NUM

print "\nOn class " + i.to_s + "..."

prior[i] = nClass[i] / nTotal

#concatenate all docs from class c

text = Array.new

trainingFiles[i.to_s].each do |dn|
    text.concat(trainingFileTerms[dn])

end

termCount = Hash.new

termCountTotal = 0.0

vocabFS.each do |t|
    termCount[t] = text.count(t).to_f
    termCountTotal = termCountTotal + termCount[t] + 1

end

vocabFS.each do |t|
    if condprob.has_key?(t) == false
        a = Array.new
        condprob[t] = a
    end

condprob[t][i] = (termCount[t] + 1) / termCountTotal

end

end
```

5. Testing

Training 結束後進入 testing phase。將 training documents 以外的其他文件一個一個打開,利用前面計算出來的條件機率,對每份文件計算其在 13 個 class 的分數,選出分數最高的 class,作為這份文件所屬的 class。

```
#for each document
| Dir.foreach( INPUT_DIR + '/') do | doc |
| next if doc == '.' or doc == '..'
| #if the document is not a training document
| if trainingDocs.include?(doc.chomp(".txt")) == false
| print "\nOn document " + doc + "..."
| f = open( INPUT_DIR + '/' + doc, 'rb').read
| terms = extract(f)
| score = Hash.new
| #compute score for each class
| for i in 1..CLASS_NUM
| score[i.to_s] = Math.log(prior[i])
| terms.each do |t|
| if vocabFS.include?(t)
| score[i.to_s] += Math.log(condprob[t][i])
| end
| end
| end
| end
| scoreSorted = score.sort_by{|key, value| value}.reverse classifyResult[doc.chomp(".txt")] = scoreSorted[0][0]
| end
| end
| output = classifyResult.sort_by{|key, value| key.to_i}
```