#### Information Retrieval and Text Mining: Homework #4

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## 執行環境

Jupyter Notebook

# 程式語言

Python 3

# 執行方式

• 在執行前需安裝 nltk 套件 (command: pip install nltk)

```
選取 Anaconda Prompt (Anaconda3) - □ ×

(base) C:\Users\user>pip install nltk

Requirement already satisfied: nltk in c:\users\user\anaconda3\lib\site-packages (3.4.5)

Requirement already satisfied: six in c:\users\user\anaconda3\lib\site-packages (from nltk) (1.12.0)

WARNING: You are using pip version 20.1.1; however, version 20.2.3 is available.

You should consider upgrading via the 'c:\users\user\anaconda3\python.exe -m pip install --upgrade pip' command.
```

- 安裝 pandas 套件 (command: pip install pandas)
- 安裝 numpy 套件 (command: pip install numpy)
- 執行環境:
  - 可以利用 Spyder 開啟 pa4.py,並執行

```
Source Console 🔻 Object
🗀 temp.py 🗵 pa4.py 🔼
         token_result = ''
         for i,token in enumerate(tokens):
             if i != len(tokens)-1:
                                                                                                                   Here you can get help of any object by pressing Ctrl+I in front of it, either on the Editor or the Console
                 token_result += ps.stem(token) + ' '
                 token_result += ps.stem(token)
                                                                                        Ē
        return(token_result)
  61 def similarity(doc1, doc2):
        d1 = pd.read_csv(doc1,names=['t_index','tf_idf'], sep=' ')
d2 = pd.read_csv(doc2,names=['t_index','tf_idf'], sep=' ')
                                                                                            Variable explorer File explorer Help
        d1 = d1.drop(0)
                                                                                            IPython console
        d2 = d2.drop(0)
                                                                                            Console 1/A 🔀
                                                                                                                                                                                            ■ # Q
                                                                                            Python 3.7.4 (default, Aug 9 2019, 18:34:13) [MSC v.1915 64 bit (AMD64)]
        d1_d2 = pd.merge(d1,d2,on='t_index', how='outer')
                                                                                            Type "copyright", "credits" or "license" for more information.
        d1_d2.fillna(0,inplace=True)
        matrix_product = np.sum(dot(d1_d2.tf_idf_x, d1_d2.tf_idf_y))
                                                                                            IPython 7.8.0 -- An enhanced Interactive Python.
        #print(matrix product
                                                                                            In [1]: runfile('C:/Users/user/R08725008/pa4.py', wdir='C:/Users/user/R08725008')
         sim = matrix_product / (norm(d1_d2.tf_idf_x)*norm(d1_d2.tf_idf_y))
                                                                                            showing info https://raw.githubusercontent.com/nltk/nltk_data/gh-pages/index.xml
  76
77 #sim = similarity('C:/Users/user/R08725008/tf_idf/1.txt','C:/Users/user/R08725008)
                                                                                            In [2]: runfile('C:/Users/user/R08725008/pa4.py', wdir='C:/Users/user/R08725008')
  78 #print(str(sim))
                                                                                            showing info https://raw.githubusercontent.com/nltk/nltk_data/gh-pages/index.xml
                                                                                            100%
                                                                                                                1095/1095 [52:11<00:00, 2.86s/it]
                                                                                                                0/1094 [00:00<?, ?it/s](1095, 1095)
                                                                                              0%
        x = np.max([C[j][i], C[j][m]]) #single link
                                                                                            100%
                                                                                                             | 1094/1094 [06:17<00:00, 2.90it/s]
        y = np.min([C[j][i], C[j][m]]) # complete link
         return (x+v)/2
```

- 確保提供的 IRTM 資料夾放於 C:/Users/user/R08725008/ 目錄下
- 確保 tf idf 資料夾放於 C:/Users/user/R08725008/ 目錄下
- 確保中間產出的 C.pkl 放於 C:/Users/user/R08725008/ 目錄下
- 產出的 8.txt, 13.txt, 20.txt 預設放於 C:/Users/user/R08725008/ 目錄下

## 作業邏輯說明

#### Part 1: Preprocessing

1.

- 利用 nltk 套件初始化 PorterStemmer
- import stopwords 套件,並增加 list,其為觀察到可能為 stopwords

```
# install NLTK
import nltk
#import string
# install related NLTK packages
nltk.download()
# Porter's algorithm
from nltk.stem.porter import *
#stopwords package
from nltk.corpus import stopwords
# to read all files in folder
```

```
# Stemming using Porter's algorithm

ps = PorterStemmer()

# Stopword lists

stop_words = set(stopwords.words('english')) #Stopword

stop_words.update(['.', ',', '"', "'", '?', '!', ':', ';', '(', ')', '[', ']', '\'', '\'', '\'', '\''', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'', '\'
```

- 2. 定義 preprocessing function,有以下功能
  - 去除 punctuation
  - 加上去除數字的處理
  - Tokenize
  - 將讀入的文件轉換為小寫,如果不在 stop words 當中的 word 才會保留
  - 最後將每個 token 進行 stemming,並加回字串當中
  - 把結果輸出 token\_result

```
def preprocessing(texts):
    texts = texts.translate(str.maketrans('', '', "!\"#$%&'()*+,-./:;<=>?@[\]^_`{|}~"))
# using translate and digits
# to remove numeric digits from string
# words in dictionary: 15144-->14348
remove_digits = str.maketrans('', '', "0123456789")
texts = texts.translate(remove_digits)

word_tokenize = texts.split()
tokens = [i.lower() for i in word_tokenize if i.lower() not in stop_words] #Stopword removal
#tokens = [i for i in tokens if not i.isdigit()]
#print(tokens)
token_result = ''
for i,token in enumerate(tokens):
    if i != len(tokens)-1:
        token_result += ps.stem(token) + ' '
    else:
        token_result += ps.stem(token)
return(token_result)
```

3. 讀入所有資料夾中的 documents,經過前處理後的所有 terms 收集至 tokens\_all

```
#read tokens in all documents

tokens_all = ""

for file in next(os.walk('C:/Users/user/R08725008/IRTM/'))[2]:
    f = open('C:/Users/user/R08725008/IRTM/'+file)
    texts = f.read()
    f.close()
    tokens_all += preprocessing(texts)
```

### Part 2: Similarity Matrix C

- 1. 定義 similarity function 計算 cosine similarity
  - 讀取 document 1 以及 document 2
  - 去除第一列 (此 document 中在 token list 中的 terms 的數量)
  - cosine similarity 分子為向量相乘,使用 numpy 中的 dot 運算
  - cosine similarity 分母為向量長度相乘,使用 numpy 中的 norm 運算 來計算向量的長度
  - 上述兩點相除,即得到 cosine similarity,此為 function 的 return value

```
In [8]:
        def similarity(doc1, doc2):
           #separate the column with space
           d1 = pd.read csv(doc1,names=['t index','tf idf'], sep=' ')
           d2 = pd.read csv(doc2,names=['t index','tf idf'], sep=' ')
           #remove the first row(counter)
           d1 = d1.drop(0)
           d2 = d2.drop(0)
           \#d2
           d1 d2 = pd.merge(d1,d2,on='t index', how='outer')
           d1 d2.fillna(0,inplace=True)
           matrix product = np.sum(dot(d1 d2.tf idf x, d1 d2.tf idf y))
           #print(matrix product)
           \#print(norm(d1 \ d2.tf \ idf \ x)*norm(d1 \ d2.tf \ idf \ y))
           sim = matrix product / (norm(d1 d2.tf idf x)*norm(d1 d2.tf idf y))
           return sim
        \#sim = similarity('C:/Users/user/R08725008/tf idf/1.txt', 'C:/Users/user/R08725008/tf idf/2.txt')
        #print(str(sim))
```

The cosine similarity of [document 1] and [document 2] is 0.23592608526919712

- 2. 將 document 做 pairwise 的 similarity 計算,以防之後會產生 zero probability 的問題,因此每個 matrix 的 entry 皆加上一個極小的數值
  - 並把計算結果的 C matrix dump 出來

```
\begin{split} N &= 1095 \\ I &= np.ones((N,), dtype=int) \\ eps &= 1e-10 \\ C &= np.zeros([N,N]) \\ \textbf{for i in } tqdm(range(N)): \\ \textbf{for j in } range(N-i-1): \\ sim &= similarity('C:/Users/user/R08725008/tf_idf/'+str(i+1)+'.txt', 'C:/Users/user/R08725008/tf_idf/'+str(j+i+2)+'.txt') \\ C[i][j+i+1] &= C[j+i+1][i] = sim + eps \\ print(C.shape) \\ pickle.dump(obj=C, file=open('C:/Users/user/R08725008/C.pkl','wb')) \end{split}
```

### Part 3: Implementing HEAP Efficiency HAC

1. 利用 single link 與 complete link 的值取平均計算 cluster 之間的 similarity

```
In [122]: def merge_sim(C, j, i, m):
    x = np.max([C[j][i], C[j][m]]) #single link
    y = np.min([C[j][i], C[j][m]]) # complete link
    return (x+y)/2

def heap_merge_sim(C, j, i, m):
    x = np.min([C[j][i][0], C[j][m][0]]) #single link
    y = np.max([C[j][i][0], C[j][m][0]]) # complete link
    return (x+y)/2
```

2. 初始化 result,一開始還沒進行合併時,每個 document 各自形成一個 cluster,並且把上一步驟所得到 C matrix load 出來使用

```
result = []
basic_result = []
C = pickle.load(open('C:/Users/user/R08725008/C.pkl','rb'))
K = [8, 13, 20]
for n in range(N):
    result.append([n])
```

- 3. 初始化 A,A 是用來記錄 merge 的過程,哪個 cluster 與哪個 cluster 做合併
  - 因為合併過程有 N-1 個 internal nodes,所以要做 N-1 次 merge
  - 找到最大的 C[i][m] 值
  - 找到後,再把 (i,m) append 至 A list

```
# basic version of HAC

A = [] # a record list of merges

for k in tqdm(range(N - 1)):

    max_sim = 0|

    max_i = 0

    max_m = 0

for i in range(N): # argmax < i,m >

    for m in range(i + 1):

    if i != m and I[i] == 1 and I[m] == 1 and C[i][m] >= max_sim:

        max_sim = C[i][m]

    max_i = i

    max_m = m

A.append((max_i, max_m))
```

- 4. 把 m cluster 合併到 i cluster, m cluster 設為 None result[max\_i] += result[max\_m] # merge m in i result[max\_m] = None
- 5. 更新其他 cluster 與 cluster i 之間的 similarity

```
for j in range(N): #update C
  the_sim = merge_sim(C, j, max_i, max_m)
  C[max_i][j] = the_sim
  C[j][max_i] = the_sim
```

6. Update I, 因為 m cluster 被合併了,再將 document 以 id 大小做排序

```
I[max_m] = 0 #update I

temp= sorted([sorted(c) for c in result if c is not None])
basic_result.append(temp)
```

7. 把分群結果 output 出來

```
# basic method of HAC
K_ = [20, 13, 8]

for k in range(len(K_)):
    with open('C:/Users/user/R08725008/'+str(K_[k])+'.txt', 'w') as f:
    for i in range(len(basic_result[k])):
        for j in range(len(basic_result[k][i])):
            f.write(str(basic_result[k][i]]+1)+'\n')
        f.write('\n')
        f.close()
```