R12725049 徐尚淵 作業二

- 1. 執行環境: Jupyter Notebook
- 2. 程式語言: Python (版本 3.11.4)

```
In [1]: from platform import python_version
    print(python_version())
3.11.4
```

3. 執行方式:

直接使用 Jupyter Notebook GUI Run code



- 4. 作業處理邏輯說明
 - 1. 先做與作業一相同的前處理 (簡化截圖)

the white house is also keeping a close watch on yugoslavia, where opposition forces are about to step up the pressure on president slobodan milosevic. but will it work? nbc's jim maceda is in belgrade tonight.

```
In [7]: # Empty List for storing processed document
processed_texts = []

for document_content in document_contents:
    result = []
    # Lower casting
    document_content = document_content.lower()
    # Tokenized
    tokenized_content = tokenize_text(document_content)
    # Stopwords removal.
    filtered_tokens = [token for token in tokenized_content if token not in stop
    # Stemming
    for t in filtered_tokens:
        result.append(ps.stem(t))
    # 荐人List
    processed_texts.append(result)

# 以文件一般測試
print(processed_texts[0])

['white', 'hous', 'also', 'keep', 'close', 'watch', 'yugoslavia', 'opposit', 'f
orc', 'step', 'pressur', 'presid', 'slobodan', 'milosev', 'work', 'nbc', 'jim',
    "maceda', 'belgrad', 'tonight', 'serbia', 'eve', 'gener', 'strike', 'two-hour',
    'roadblock', 'tast', 'come', 'tomorrow', 'say', 'opposit', 'nationwid', 'work',
```

2. 計算每個文字在 Document 當中出現的頻率,利用 term_info_List 儲存之後可能會用到的資訊。

```
In [8]:
# Empty list 用以儲存每個量字的index, term, frequency
term_info_list = []

# Empty dict for storing term_document_frequency
term_document_frequency = {}

# Loop start
for words in processed_texts:
# 接免重複計算
    unique_words = set(words)

# 更新Frequency
for word in unique_words:
    if word in term_document_frequency:
        term_document_frequency[word] += 1
    else:
        term_document_frequency[word] = 1

# 按照字母順序排列字典
sorted_terms = sorted(term_document_frequency.items(), key-lambda x: x[0])

# 紀錄 index 並存人 term_info_list 中
for index, (term, frequency) in enumerate(sorted_terms, start=1):
    term_info = { 'index': index, 'term': term, 'frequency': frequency}
term_info_list.append(term_info)
term_info_list.append(term_info)
```

3. 將 Dictionary 輸出並命名為 dictionary.txt

```
:# 顧存字典為txt
output_file = "dictionary.txt"
with open(output_file, "w", encoding="utf-8") as file:
    # Header
    file.write("t_index term df\n")
    for term_info in term_info_list:
        line = "{:<8} {:<25} {}\n".format(term_info['index'], term_info['term'])
        file.write(line)
print(f"字典已保存到 {output_file}")

字典已保存到 dictionary.txt</pre>
```

4. 計算每個文件的 tf-idf vector

```
: # 以随機文件做阅新
print(tfidf_vectors[597])
[{'index': 14237, 'tf-idf': 0.0009284568811649149, 'tf': 0.00263852242744063
3, 'idf': 0.35188515796150277}, ('index': 10994, 'tf-idf': 0.0380854911796817
```

5. 將每個文件的 tf-idf 儲存

```
output_folder = "output"
os.makedirs(output_folder, exist_ok=True)

# 將每國文件的 tf-idf 顧存
for document_index, words in enumerate(processed_texts, start=1):
# 文件名為 Doctd.txt
document_filename = f"{document_index}.txt"

with open(os.path.join(output_folder, document_filename), "w", encoding="utf # Header
    file.write("t_index tf-idf\n")

for tfidf_info in tfidf_vectors[document_index - 1]:
    line = "{:<8} {}\n".format(tfidf_info['index'], tfidf_info['tf-idf' file.write(line)
```

6. Cosine Similarity Function

```
In [13]:

def cosine(vector_x, vector_y):
# 計算向量大榜

dot_product = sum(x['tf-idf'] * y['tf-idf'] for x, y in zip(vector_x, vector_y)

# 計算向量大小

magnitude_x = math.sqrt(sum(x['tf-idf']**2 for x in vector_y))

# Cosine similarity

similarity = dot_product / (magnitude_x * magnitude_y)

return similarity

def unit_vector(vector):
# 計算向量大小

magnitude = math.sqrt(sum(x['tf-idf']**2 for x in vector))

# 計算單位向量

unit_vector = [{'tf-idf': x['tf-idf'] / magnitude} for x in vector]

return unit_vector
```

7. 先計算單位向量再計算 Cosine Similarity

```
In [14]: # 計算兩文件各自的單位向量
unit_vector_x = unit_vector(tfidf_vectors[154])
unit_vector_y = unit_vector(tfidf_vectors[958])

# 計算cosine similarity
similarity = cosine(unit_vector_x, unit_vector_y)

print(f"文件 x 和文件 y 的cosine similarity: {similarity}")

文件 x 和文件 y 的cosine similarity: 0.6642675123400708
```