Lab 1 - Building dictionary with Selma Lagerlöf novels

Laboration 1 in EDAN20 @ LTH - http://cs.lth.se/edan20/coursework/assignment-1/_(http://cs.lth.se/edan20/coursework/assignment-1/)

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The objectives of this assignment are to:

-Write a program that collects all the words from a set of documents

-Represent a document using the Tf.ldf value

-Write a short report of 1 to 2 pages on the assignment

-Read a short text on an industrial system

Indexing one file

- The index file will contain all the unique words in the document, where each word is associated with the list of its positions in the document.
 You will represent this index as a dictionary where the keys will be the words and the values, the lists of positions
 As words, you will consider all the strings of letters that you will set in lower case. You will not index the rest (i.e. numbers or symbols).
 To extract the words, you will use Unicode regular expressions. Do not use \(\text{W+}, \) for instance, but the Unicode equivalent. The word positions will correspond to the number of characters from the beginning of the file. (The word offset from the beginning)
 You will use finditer() to find the positions of the words. This will return you match objects, where you will get the matches and the positions with the group() and start() methods.

```
In [1]: text = open("Selma/bannlyst.txt").read()
In [2]: import regex as re
              Helper function
              return matchobj.group(1).lower()
             Make text lowercase and put into list
              textLow = re.sub(r'(\p{Lu})', toLowercase, text) # Lowercase all characters
             stringList = re.findall(r"\p{L}+",textLow) # This finds all words from a txt file. r"[a-zåäö]+ equal to r"\w+"
In [3]: def toLowercase(matchobj):
    return matchobj.group(1).lower()
          def string2dict(text):
             Creates a dict with (word:list[index apperences]) from input string
               textLow = re.sub(r'(\p{Lu})', toLowercase, text) # Lowercase all characters stringList = textLowerList(text) stringDist = (key : list([) for key in stringList)
              for m in re.finditer(r"\p{L}+", textLow): # Iterate thorugh every word
                   s = m.start()
e = m.end()
                   word = textLow[s:e]
stringDict[word].append(s)
              return stringDict
In [4]: txtDict = string2dict(text)
In [5]: len(txtDict)
Out[5]: 7923
```

The word gjord occurs three times in the text at positions 8551, 183692, and 220875, uppklarnande, once at position 8567, and stjärnor, once at position 8590

```
In [6]: txtDict['gjord']
Out[6]: [8551, 183692, 220875]
In [7]: txtDict['uppklarnande']
Out[7]: [8567]
In [8]: txtDict['stjärnor']
Out[8]: [8590]
```

You will use the pickle package to write your dictionary in an file, see https://wiki.python.org/moin/UsingPickle (https://wiki.python.org/moin/UsingPickle)

```
In [9]: import pickle
             #with open('BannlystTxtDict.pickle', 'wb') as handle:
# pickle.dump(txtDict, handle, protocol=pickle.HIGHEST PROTOCOL)
```

Test of pickle

```
In [11]: BannlystTxtDict == txtDict
```

Reading the content of a folder

Use function:

```
In [12]: import os
              def get_files(fileDir, suffix):
                   Returns all the files in a folder ending with suffix ;param filedir:
                    :param suffix:
:return: the list of file names
                  files = []
for file in os.listdir(fileDir):
    if file.endswith(suffix):
        files.append(file)
return files
In [13]: files = get_files("selma", ".txt")
In [14]: files
'jerusalem.txt',
'bannlyst.txt',
'gosta.txt']
```

Creating a master index

Complete your program with the creation of master index, where you will associate each word of the corpus with the files, where it occur and its positions. (a posting list)

```
In [15]: def toLowercase(matchobj):
    return matchobj.group(1).lower()
             def addAll(fileDir,files):
                 This function takes way to long. Do not iterate word in dict but build dict directly.
                  Reads all files in list and matches to txt files
                 totText = []
                 for file in files:
    text = open(fileDir+"/"+file).read()
    stringList = textLowerList(text)
    totText.extend(stringList)
                 masterDict = {word : {file : list([]) for file in files} for word in totText}
                for file in files:
	text = open(fileDir+"/"+file).read()
	txtDict = stringZdict(text)
	for word in txtDict.keys():
	masterDict[word][file] = txtDict[word]
                 return masterDict
In [16]: masterDict = addAll('selma',files)
In [18]: with open('masterDict.pickle', 'rb') as handle:
    masterDict = pickle.load(handle)
```

Test of master dict. Below is an except of the master index with the words samlar and ände:

'samlar': {'nils.txt': [53499, 120336], 'gosta.txt': [317119, 414300, 543686], 'osynliga.txt': [410995, 871322]}

```
In [19]: masterDict["samlar"]
Out[19]: {\text{'troll.txt': [], \text{'kejsaren.txt': [], \text{'marbacka.txt': [], \text{'herrgard.txt': [], \text{'niis.txt': [53499, 120336], \text{'osynliga.txt': [41095, 871322], \text{'jerusalem.txt': [], \text{'banhyst.txt': [], \text{'gosta.txt': [317119, 414300, 543686]}}
```

'ände':{'nils.txt': [3991],'kejsaren.txt': [51100],'marbacka.txt': [374231],'troll.txt': [39726],'osynliga.txt': [742747]},

```
In [20]: | AmasterDict[ and a ]
Out[20]: | AmasterDict[ and a ]
Nejsaren.txt': [51100],
'marbacka.txt': [374231],
'herrgard.txt': [7,
'nils.txt': [3991],
'osynliga.txt': [742747],
'jerusalem.txt': [],
'bannlyst.txt': [],
'gosta.txt': [])
```

Representing Documents with tf-idf

Once you have created the index, you will represent each document in your corpus as a word vector. You will define the value of a word in a document with the tf-idf metric. Tf will be the relative frequency of the term in the document and idf, the logarithm base 10 of the inverse

```
In [21]: import math
                      Creates a ft-idf dict from all files.
https://www.freecodecamp.org/news/how-to-process-textual-data-using-tf-idf-in-python-cd2bbc0a94a3/
:return dict:
                     This function takes way to long. Do not iterate word in dict but build dict directly.
                     tfIdfDict = masterDict.copy()
                     lenText = {}
for file in masterDict['nils']: # Read total nbr of words in each text
    text = open('selma'+')"-file).read()
    lenText[file] = len(textLowerList(text)) # nbr of words in textfile
                     for word in masterDict:
    #idf will be the logarithm base 10 of the inverse document frequency.
    nbrKeys = len(masterDict(word).keys())
    dictValues = masterDict(word).values()
    lenDictValues = len(dictValues)
                           for fileList in dictValues: # Count nbr of empty list. (There is probably a better way to do this) if not fileList: i = i + 1
                           df = (lenDictValues-i)
                           idf = math.log10(nbrKeys/df)
                           for file in masterDict[word]:
    # Tf will be the relative frequency of the term in the document
    lenWordVec = len(masterDict[word][file]) # nbr of occurencies of word
                                 tf = lenWordVec / lenText[file]
tfIdfDict[word][file] = tf*idf
                     return tfIdfDict
 In [22]: tfIdfDict = tiIdf(masterDict)
 Test of tf idf:
känna :: bannlyst.txt 0.0, gosta.txt 0.0, herrgard.txt 0.0, jerusalem.txt 0.0, nils.txt 0.0
 In [25]: tfIdfDict['känna']
gås :: bannlyst.txt 0.0, gosta.txt 0.0, herrgard.txt 0.0, jerusalem.txt 0.0, nils.txt 0.00010123719421964931
 In [26]: tfIdfDict['gås']
 'marbacka.txt': 0.0,
'herrgard.txt': 0.0,
'hils.txt': 0.001012371942196493,
'osynliga.txt': 0.0,
'jerusalem.txt': 0.0,
'bannlyst.txt': 0.0,
'gosta.txt': 0.0,
nils :: bannlyst.txt 0.0, gosta.txt 0.0, herrgard.txt 0.0 jerusalem.txt 4.778415355159037e-06, nils.txt 9.801209641132888e-05
 In [27]: tfIdfDict['nils']
et :: bannlyst.txt 6.2846093167673765e-06, gosta.txt 0.0, herrgard.txt 0.0, jerusalem.txt 0.0, nils.txt 0.0
 In [28]: tfIdfDict['et']
In [28]: tridfbzt('et']
Out[28]: {'troll.txt': 0.0,
    'kejsaren.txt': 6.0441957178149187e-05,
    'marbacka.txt': 1.0.4187791927195648e-05,
    'herrgard.txt': 0.0,
    'osynliga.txt': 0.0,
    'jerusalem.txt': 0.0,
    'bannlyst.txt': 6.2846093167673765e-06,
    'gosta.txt': 0.0}
Comparing Documents
Using the cosine similarity, compare all the pairs of documents with their tfidf representation and present your results in a matrix. You will include this matrix in your report.
Give the name of the two novels that are the most similar
There are the document representations in term of words. Rows: documents, Col: words.
 In [29]: import numpy as np
               docMatrix = np.zeros((9,len(tfIdfDict.keys())))
wordList = tfIdfDict.keys()
fileList = tfIdfDict['nils']
               for i, word in enumerate(wordList):
    for j, file in enumerate(fileList):
        docMatrix[j,i] = tfIdfDict[word][file]
```

In [30]: import numpy as np; import pandas as pd from sklearn.metrics.pairwise import cosine_similarity

df = pd.DataFrame(docMatrix)
simularityMatrix = cosine similarity(df)

```
In [32]: import matplotlib.pvplot as plt
                                                                 plt.imshow(np.log(simularityMatrix))
                                                               for i,file in enumerate(list(fileList.keys())):
    print(i, "=", file)
                                                           0 = troll.txt

1 = kejsaren.txt

2 = marbacka.txt

3 = herrgard.txt

4 = nils.txt

5 = osynliga.txt

6 = jerusalem.txt

7 = bannlyst.txt

8 = gosta.txt
 In [33]: print("TEXT 1", "TEXT 2", "Simularity Value")
for i,iFile in enumerate(list(fileList.keys())):
                                                                                         for j, jFile in enumerate(list(fileList.keys()
    print(iFile, jFile, simularityMatrix[i,j])
                                                         print(iFile, jFile, simularityMatris)

TEXT 1 TEXT 2 Simularity Value

troll.txt kejsaren.txt 0.0883126344462

troll.txt kejsaren.txt 0.0883126344462

troll.txt marback.txt 0.0236485882322

troll.txt mis.txt 0.0133870728052

troll.txt mis.txt 0.0133870728052

troll.txt mis.txt 0.0133870728052

troll.txt mis.txt 0.007362425230154

troll.txt jerusalem.txt 0.00706370160499

troll.txt jerusalem.txt 0.00736370160499

troll.txt jerusalem.txt 1.00883126344462

kejsaren.txt troll.txt 0.0883126344462

kejsaren.txt troll.txt 0.0883126344462

kejsaren.txt troll.txt 0.088312634462

kejsaren.txt marbacka.txt 0.044734652002

kejsaren.txt inis.txt 0.044734652002

kejsaren.txt jerusalem.txt 0.00509834658

kejsaren.txt jerusalem.txt 0.00509834658

marbacka.txt troll.txt 0.00509834658

marbacka.txt troll.txt 0.0256886882832

marbacka.txt troll.txt 0.0407146550004

marbacka.txt marbacka.txt 1.0

marbacka.txt marbacka.txt 1.0

marbacka.txt injerusalem.txt 0.04079927496

marbacka.txt injerusalem.txt 0.012904030053

marbacka.txt troll.txt 0.00250983409828077

marbacka.txt tosynliga.txt 0.005099349892077

marbacka.txt tosynliga.txt 0.005099349892077

marbacka.txt tosynliga.txt 0.005099349892077

marbacka.txt tosynliga.txt 0.005099349892077

marbacka.txt tosynliga.txt 0.005099349893077

marbacka.txt tosynliga.txt 0.005099349893077

marbacka.txt tosynliga.txt 0.005099349893077

marbacka.txt tosnliga.txt 0.00398629497309

herrgard.txt kejsaren.txt 0.0039862947309

herrgard.txt kejsaren.txt 0.0039862947309

herrgard.txt marbacka.txt 0.0039862947309

herrgard.txt marbacka.txt 0.0019879927496
                                                                   TEXT 1 TEXT 2 Simularity Value
                                                         marbacka.txt jerusalem.txt 0.012904309203
marbacka.txt banniyst.txt 0.0509649828077
marbacka.txt gosta.txt 0.0259649828077
marbacka.txt tyosta.txt 0.003986298428079
mergard.txt troll.txt 0.003986298428921853
herrgard.txt kejsaren.txt 0.000892488921853
herrgard.txt misl.sxt 0.0114169936563
herrgard.txt injls.txt 0.0114169936563
herrgard.txt jerusalem.txt 0.007215128127
herrgard.txt banniyst.txt 0.0125736366664
herrgard.txt banniyst.txt 0.00129900476994
herrgard.txt banniyst.txt 0.00129900476994
herrgard.txt bout 0.0043743662002
nils.txt troll.txt 0.0141469935566
nils.txt troll.txt 0.0141469935663
nils.txt instragard.txt 0.0402145502026
nils.txt herrgard.txt 0.0414169336563
nils.txt instragard.txt 0.0321284846367
nils.txt perusalem.txt 0.013097803341
nils.txt perusalem.txt 0.013097803341
nils.txt perusalem.txt 0.013097803341
nils.txt perusalem.txt 0.01059434358028
osynliga.txt troll.txt 0.0292661220154
osynliga.txt troll.txt 0.032128484637
osynliga.txt injls.txt 0.03212848637
osynliga.txt injls.txt 0.03212848637
osynliga.txt injls.txt 0.03212848637
osynliga.txt perusalem.txt 0.0418440605847
osynliga.txt perusalem.txt 0.018978033341
perusalem.txt troll.txt 0.00706371016099
perusalem.txt troll.txt 0.00706371016099
perusalem.txt troll.txt 0.00706371016099
perusalem.txt troll.txt 0.00706371016099
perusalem.txt troll.txt 0.0070872033011
perusalem.txt troll.txt 0.0070878033341
perusalem.txt troll.txt 0.00708780733341
perusalem.txt troll.txt 0.0073840636582
postal.txt harbacka.txt 0.0082987707788
postal.txt he
In [34]: index = simularityMatrix[:,:].flatten().argsort()[-10:][::-1][9] # This is the
                                                                 print("Position: (", index//9, ",", index, ") has maximum val")
                                                               print("Closest documents are:", list(fileList.keys())[index//9], "and", list(fileList.keys())[index], ". With cosine value:", simularityMatrix[0,1])
```

Reading

Position: (0 , 1) has maximum val Closest documents are: troll.txt and kejsaren.txt . With cosine value: 0.0883126344462

Read the text: Challenges in Building Large-Scale Information Retrieval Systems about the history of Google indexing by Jeff Dean. In your report, tell how your index encoding is related to what Google did. You must identify the slide where you have the most similar indexing technique and write the slide number in your report. https://static.googleusercontent.com/media/research.google_com/en//people/jeff/WSDM09-keynote.pdf/

Answer

Google had a number of index shards (words). The search time to find the requested sharde could be improved by improving index encoding. Similar to our encoding google searched for a shard (word) to find attributes for that word. In our case file namne + tfidx value is the attribute for each word, but in google case font size, title, etc. This is explained on page 45.