GenomikOdev2(031690058-031790058)

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1 GENOMİK HESAPLAMA ÖDEVİ 2 (31.05.2021)

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```
[1]: # gerekli kütüphaneler import edilir

import pandas as pd
import numpy as np
import nltk

nltk.download("stopwords")
from nltk.corpus import stopwords

nltk.download("wordnet")
from nltk.stem import WordNetLemmatizer

from nltk.stem import PorterStemmer
```

```
[2]: addr = r"C:\Users\Asus\Desktop\compGenomic_hw2\texts\dataset\*txt"

corpus = []

import glob
import errno

files = glob.glob(addr)
for name in files:
    try:
```

```
with open(name) as f:
    raw_text = f.read()
    corpus.append(raw_text)
    f.close()
    except IOError as exc:
    if exc.errno != errno.EISDIR:
        raise
#print(corpus)
# metinleri okuyup diziye attik
```

[]: corpus[9]

```
[3]: # okunan metinlerde nlp işlemleri uygula: kelimeleri tokenla, noktalama
     →işaretlerini kaldıe ve köklerine ayır
    corpus_processed = []
    for i in range(len(corpus)):
        temp_text = corpus[i]
         #print("\n ******* \n")
         #print(temp_text)
        tokenList = nltk.word_tokenize(temp_text)
        tokenList = list(filter(lambda token: nltk.tokenize.punkt.PunktToken(token).
     →is_non_punct, tokenList))
        tokenList = list(filter(lambda token: token not in stopwords.
     ⇔words("english"), tokenList))
         #lemmatizer = WordNetLemmatizer()
         #tokenList = [lemmatizer.lemmatize(word) for word in tokenList]
         stemmer = PorterStemmer()
        tokenList = [stemmer.stem(word) for word in tokenList]
        tokenList = [word.lower() for word in tokenList]
         corpus_processed.append(tokenList)
```

[]: print(corpus_processed)

```
[]: corpus_processed[1]
```

```
[4]: txts = ["A1", "A2", "A3", "A4", "A5", "A6", "A7", "A8", "A9", "A10"]
#index (0, 1, 2, 3, 4, 5, 6, 7, 8, 9)
```

```
for x in range(len(corpus_processed)):
             tokens = corpus_processed[x]
             txts[x] = pd.DataFrame(tokens, columns = ["Words"])
             txts[x].sort_values("Words", ascending = True, inplace = True,
      →ignore_index=True)
[5]: pd.options.display.max_rows = 100
[6]: txts[0] # sadece ilk metni yazdırdık, txts dizisi tüm metin tablolarını tutar
[6]:
            Words
     0
             1862
     1
             1887
     2
            2,500
     3
          academi
          academi
     203
            vault
    204
              way
     205
             wine
     206
             work
     207
             work
     [208 rows x 1 columns]
[7]: # her metin için ayrı ayrı tf hesapla
     for x in range(len(txts)):
         df = txts[x]
         term_freq = np.zeros(len(df))
         term_idf = np.zeros(len(df))
         df["word_count"] = term_freq
         df["tf"] = term_idf
         count = 0
         for i in range(len(df)):
             word = df["Words"].values[i]
             #count += 1
             #j = i+1
             for j in range(len(df)):
                 nextWord = df["Words"].values[j]
                 if(word == nextWord):
                     count +=1
                 #else:
```

```
df["word_count"].values[i] = count # kelimenin bir metinde kaç kez_
       →geçtiği bilgisidir
              df["tf"].values[i] = count / len(df) # kelimenin geçme sayısı/
       →metindeki tüm kelimelerin sayısı
              count = 0
          df.drop_duplicates(inplace=True, ignore_index=True) # tablonun quzelu
       →gözükmesi için tekrarlayan satırları sildik
          # silme işlemi tablonun uzunluğunu bastırırken etkiler, tf hesaplarken
       \rightarrow etkilemez
          df.set_index(np.arange(len(df)), inplace = True)
          txts[x] = df
          #print(x)
 [8]: txts[0]
 [8]:
             Words word_count
              1862
                           1.0 0.004808
      0
      1
              1887
                           1.0 0.004808
      2
             2,500
                           1.0 0.004808
      3
                           2.0 0.009615
           academi
              acid
                           1.0 0.004808
                           1.0 0.004808
      154
           variou
      155
            vault
                           1.0 0.004808
                           1.0 0.004808
      156
               way
      157
              wine
                           1.0 0.004808
      158
              work
                           2.0 0.009615
      [159 rows x 3 columns]
[10]: # her kelime için idf hesapla
      for x in range(len(txts)):
          df = txts[x]
          doc_freq = np.zeros(len(df))
          df["idf"] = doc_freq
          doc_num = np.zeros(len(df))
          df["num_of_doc"] = doc_num
          count = 0
```

for i in range(len(df)): #dfin içindeyiz kelime seçiyoruz

#df (metin) seçtik

word = df["Words"].values[i]

```
for j in range(len(txts)): # diğer dfleri kontrol edecek o kelime varu
→mr diye
           ddf = txts[j]
           if(word in ddf["Words"].values):
               count +=1
       df["num_of_doc"].values[i] = count
       df["idf"].values[i] = round(10/count,2) # toplamda 10 adet doküman vardı
       count = 0
  txts[x] = df
   #print(x)
```

[11]: txts[0]

```
[11]:
            Words
                   word_count
                                           idf num_of_doc
                                      tf
      0
             1862
                           1.0 0.004808
                                           5.0
                                                       2.0
      1
             1887
                           1.0 0.004808 10.0
                                                       1.0
      2
             2,500
                           1.0 0.004808 10.0
                                                       1.0
      3
           academi
                           2.0 0.009615
                                         10.0
                                                       1.0
                           1.0 0.004808
                                          5.0
      4
             acid
                                                       2.0
               •••
                           1.0 0.004808
                                        10.0
                                                       1.0
      154
           variou
      155
            vault
                           1.0 0.004808
                                         5.0
                                                       2.0
                                                       1.0
      156
                           1.0 0.004808 10.0
               way
                                                       1.0
      157
                           1.0 0.004808 10.0
             wine
                                                      10.0
      158
             work
                           2.0 0.009615
                                         1.0
```

[159 rows x 5 columns]

```
[]: A = 0 # tüm metin tabloları yazdırır
     for i in range(len(txts)):
        A = A+1
        print("\n ***** \n")
        print(" Text {} : ".format(A))
        print(txts[i])
```

1.0.1 İSTENEN 2 : A1 ile A3 tf-idf değerlerini hesaplayın

```
[12]: A1 = txts[0]
      A1
```

```
[12]:
            Words word_count
                                      tf
                                           idf num_of_doc
      0
              1862
                           1.0 0.004808
                                           5.0
                                                       2.0
      1
              1887
                           1.0 0.004808 10.0
                                                       1.0
```

```
2
       2,500
                     1.0 0.004808 10.0
                                                 1.0
3
     academi
                     2.0 0.009615
                                   10.0
                                                 1.0
                                     5.0
                                                 2.0
4
        acid
                     1.0 0.004808
. .
154
     variou
                     1.0 0.004808
                                   10.0
                                                 1.0
155
                     1.0 0.004808
                                     5.0
                                                 2.0
      vault
                     1.0 0.004808 10.0
                                                 1.0
156
         way
157
                     1.0 0.004808 10.0
                                                 1.0
        wine
158
                     2.0 0.009615
                                                10.0
                                    1.0
        work
```

[159 rows x 5 columns]

```
[13]: A3 = txts[2]
A3
```

| [13]: | Words | word_count | tf | idf | num_of_doc |
|-------|---------|------------|----------|-------|------------|
| 0 | 's | 9.0 | 0.019780 | 1.11 | 9.0 |
| 1 | 10 | 1.0 | 0.002198 | 2.50 | 4.0 |
| 2 | 17 | 1.0 | 0.002198 | 5.00 | 2.0 |
| 3 | 1864 | 1.0 | 0.002198 | 10.00 | 1.0 |
| 4 | 1867 | 1.0 | 0.002198 | 10.00 | 1.0 |
| | | ••• | | | ••• |
| 300 | york | 1.0 | 0.002198 | 10.00 | 1.0 |
| 301 | young | 1.0 | 0.002198 | 10.00 | 1.0 |
| 302 | youth | 1.0 | 0.002198 | 10.00 | 1.0 |
| 303 | zweiten | 1.0 | 0.002198 | 10.00 | 1.0 |
| 304 | über | 1.0 | 0.002198 | 5.00 | 2.0 |

[305 rows x 5 columns]

2 Inverted Index Oluşturun

```
[14]: #concat -> inverted index oluştur, metin tablolarını birleştirerek

inv_index = pd.DataFrame()

for df in range(len(txts)):
    x = txts[df].copy()
    x.drop(columns = ["tf"], inplace = True)
    inv_index = pd.concat([inv_index, x], axis = 0, ignore_index=True)

inv_index
```

```
[14]: Words word_count idf num_of_doc
0 1862 1.0 5.00 2.0
1 1887 1.0 10.00 1.0
```

```
2
           2,500
                           1.0 10.00
                                               1.0
3
         academi
                           2.0
                               10.00
                                               1.0
4
                                 5.00
            acid
                           1.0
                                               2.0
                           •••
2202
                           1.0
                                 1.00
                                              10.0
            work
2203
                           1.0
                               10.00
                                               1.0
      world-lead
2204
         written
                           1.0
                                10.00
                                               1.0
2205
                                 1.25
                                               8.0
            year
                           3.0
2206
                                               1.0
       zoologist
                           1.0 10.00
[2207 rows x 4 columns]
```

```
[15]: cols = ["A1", "A2", "A3", "A4", "A5", "A6", "A7", "A8", "A9", "A10"]

for i in range(len(cols)):
    col_name = cols[i]
    col = np.zeros(len(inv_index))
    inv_index[col_name] = col
    #print(col_name)
```

```
[16]: inv_index # boş sütunlar eklendi
```

```
[16]:
                Words
                       word_count
                                     idf
                                          num_of_doc
                                                            A2
                                                                                A6
                                                       Α1
                                                                 A3
                                                                      A4
                                                                           A5
                                    5.00
     0
                 1862
                              1.0
                                                 2.0 0.0
                                                           0.0
                                                                0.0
                                                                     0.0
                                                                          0.0
                                                                               0.0
     1
                 1887
                              1.0
                                  10.00
                                                 1.0 0.0 0.0
                                                               0.0
                                                                     0.0
                                                                               0.0
                                                                          0.0
     2
                2,500
                              1.0
                                   10.00
                                                 1.0 0.0 0.0
                                                                0.0
                                                                     0.0
                                                                          0.0 0.0
     3
                                                 1.0 0.0
                                                           0.0
                                                                0.0
                                                                     0.0
                                                                          0.0
                                                                               0.0
              academi
                              2.0
                                   10.00
                                    5.00
                                                 2.0 0.0
                                                           0.0
                                                                0.0
                                                                         0.0 0.0
     4
                 acid
                              1.0
                                                                     0.0
                               •••
     2202
                              1.0
                                    1.00
                                                10.0
                                                      0.0
                                                           0.0
                                                               0.0
                                                                     0.0 0.0 0.0
                 work
     2203 world-lead
                              1.0
                                  10.00
                                                 1.0 0.0 0.0
                                                               0.0
                                                                     0.0 0.0 0.0
     2204
              written
                                   10.00
                                                 1.0 0.0 0.0
                                                                0.0
                                                                     0.0 0.0 0.0
                              1.0
     2205
                                                                0.0
                              3.0
                                    1.25
                                                 8.0 0.0 0.0
                                                                     0.0 0.0 0.0
                 year
     2206
            zoologist
                              1.0 10.00
                                                 1.0 0.0 0.0
                                                               0.0 0.0 0.0 0.0
```

```
0
     0.0
          0.0
               0.0
                    0.0
     0.0
         0.0
               0.0
                    0.0
1
2
     0.0 0.0
               0.0
                    0.0
3
     0.0 0.0
               0.0
                    0.0
     0.0 0.0
               0.0
                    0.0
2202 0.0 0.0
               0.0
                    0.0
2203
     0.0 0.0
               0.0
                    0.0
2204 0.0 0.0
               0.0
                    0.0
2205
     0.0 0.0
               0.0
                    0.0
2206 0.0 0.0
               0.0 0.0
```

8A

Α9

A10

A7

[2207 rows x 14 columns]

```
[17]: # her kelime için inverted index oluştur
     for i in range(len(inv_index)): #dfin içindeyiz kelime seçiyoruz
         word = inv index["Words"].values[i]
         cur_col = 4
         for j in range(len(txts)): # diğer dfleri kontrol edecek o kelime var mı⊔
      \rightarrow diye
             ddf = txts[j]
             ccol = inv_index.columns[cur_col]
             if(word in ddf["Words"].values):
                 inv index[ccol].values[i] = 1
                 #print(ccol)
             cur_col += 1
[18]: inv_index.drop(["word_count"], axis=1, inplace=True)
     inv_index.drop(["idf"], axis=1, inplace=True) # bu iki bilqi inverted index'te_
      →gerekli değil
     inv index
      # kelimenin bulunduğu dokümanlar 1, olmayanlar 0 değerini alır
[18]:
                Words
                       num_of_doc
                                    Α1
                                         A2
                                              ΑЗ
                                                  Α4
                                                       A5
                                                            A6
                                                                 A7
                                                                      8A
                                                                           Α9
                                                                               A10
                                  1.0 0.0 0.0
                                                           0.0 1.0 0.0
     0
                 1862
                                                 0.0
                                                      0.0
                                                                          0.0
                                                                               0.0
                              2.0
     1
                 1887
                              1.0 1.0 0.0 0.0
                                                 0.0
                                                      0.0
                                                           0.0
                                                                0.0 0.0
                                                                          0.0
                                                                               0.0
     2
                2,500
                              1.0 1.0
                                        0.0 0.0 0.0
                                                      0.0
                                                           0.0 0.0 0.0
                                                                          0.0
                                                                               0.0
     3
              academi
                              1.0 1.0
                                        0.0
                                            0.0
                                                 0.0
                                                      0.0
                                                           0.0 0.0
                                                                     0.0
                                                                          0.0
                                                                               0.0
     4
                 acid
                              2.0 1.0 0.0 0.0 0.0
                                                      0.0
                                                           0.0 0.0
                                                                     0.0
                                                                          0.0
                                                                               1.0
     2202
                                  1.0 1.0 1.0
                                                 1.0 1.0
                                                           1.0 1.0 1.0
                             10.0
                                                                          1.0
                                                                               1.0
                 work
     2203 world-lead
                                                           0.0 0.0 0.0
                                                                               1.0
                              1.0 0.0 0.0 0.0 0.0
                                                      0.0
                                                                          0.0
     2204
              written
                              1.0 0.0 0.0 0.0 0.0
                                                      0.0
                                                           0.0 0.0 0.0
                                                                          0.0
                                                                               1.0
     2205
                 vear
                              8.0 0.0 1.0 1.0 1.0
                                                      1.0
                                                           1.0 1.0 0.0
                                                                          1.0
                                                                               1.0
     2206
                              1.0 0.0 0.0 0.0
                                                 0.0
                                                      0.0
                                                           0.0 0.0 0.0
            zoologist
                                                                          0.0
                                                                               1.0
     [2207 rows x 12 columns]
```

3 Sorgu metni (A11) en çok hangi metne benziyor

```
[19]: # sorgu metni okunur ve aynı nlp işlemlerinden geçer
      query addr = r"C:\Users\Asus\Desktop\compGenomic hw2\texts\query\A11.txt"
      with open(query_addr) as f:
              query_text = f.read()
              tokens = nltk.word_tokenize(query_text)
              tokens = list(filter(lambda token: nltk.tokenize.punkt.
       →PunktToken(token).is_non_punct, tokens))
              tokens = list(filter(lambda token: token not in stopwords.
       →words("english"), tokens))
              lemmatizer = WordNetLemmatizer()
              tokens = [lemmatizer.lemmatize(word) for word in tokens]
              stemmer = PorterStemmer()
              tokens = [stemmer.stem(word) for word in tokens]
              tokens = [word.lower() for word in tokens]
              f.close()
      query df = pd.DataFrame(tokens, columns = ["words"])
      query_df.sort_values("words", ascending = True, inplace = True,
      →ignore_index=True)
      query_df
```

[19]: words

- 0 comput
- 1 physic
- 2 scienc
- 3 theori

3.1 Metinlerin benzerlik oranı aşağıdaki formülle hesaplanmıştır

(bkz. Introduction to Probabilistic Models for Information Retrieval - Victor Lavrenko / University of Edinburgh)

```
[20]: # A11 en çok hangisine benziyor bulunur

doc_scores = {} # tüm metinlerin A11'e benzerlik puanını tutar

total_score = 1
  max_score = 0
  N = 10 # number of documents
  constant = 0.5
```

```
for x in range(len(txts)): # df seçtik
    df = txts[x]
    cur_doc = cols[x]
    for i in range(len(df)): #dfin içindeyiz kelime seçiyoruz
        word = df["Words"].values[i]
          # diğer dfleri kontrol edecek o kelime var mı diye
        if(word in query_df["words"].values):
            print("text is:", cur_doc ,"word is:", word,"location is:", i)
            word_docnum = df["num_of_doc"].values[i]
            word_score = float((N - word_docnum + constant) / word_docnum +
 →constant)
            total_score *= word_score
             #print("word score is", word_score)
    if(total_score == 1):
        total_score = 0 # eğer iki metin ortak kelime barındırmıyorsa tatalu
 \rightarrowskor 0
        print("ortak kelime yok")
    total_score = round(total_score,2)
    doc_scores[cur_doc] = total_score
    print("similarity score is :", total_score)
    print("\n")
    if(total_score > max_score):
        max_score = total_score
    total_score = 1 # farklı bir metne geçerken 1'e eşitlenir
text is: A1 word is: scienc location is: 134
text is: A1 word is: theori location is: 145
similarity score is : 2.56
text is: A2 word is: physic location is: 101
text is: A2 word is: theori location is: 138
similarity score is: 3.4
text is: A3 word is: physic location is: 195
text is: A3 word is: theori location is: 258
similarity score is: 3.4
```

```
ortak kelime yok
     similarity score is : 0
     text is: A5 word is: comput location is: 78
     text is: A5 word is: scienc location is: 199
     text is: A5 word is: theori location is: 221
     similarity score is: 7.68
     text is: A6 word is: comput location is: 50
     text is: A6 word is: scienc location is: 186
     text is: A6 word is: theori location is: 208
     similarity score is: 7.68
     ortak kelime yok
     similarity score is : 0
     text is: A8 word is: comput location is: 39
     text is: A8 word is: physic location is: 107
     similarity score is: 6.38
     text is: A9 word is: physic location is: 179
     text is: A9 word is: scienc location is: 211
     similarity score is: 3.4
     text is: A10 word is: scienc location is: 135
     similarity score is : 1.6
[21]: doc_scores # görülüyor ki sorgu metni (A11) en çok A5 ve A6'ya benziyor
[21]: {'A1': 2.56,
       'A2': 3.4,
       'A3': 3.4,
       'A4': 0,
       'A5': 7.68,
       'A6': 7.68,
       'A7': 0,
       'A8': 6.38,
       'A9': 3.4,
       'A10': 1.6}
```