Assignment 2

Hussam Hallak

CS834, Information Retrieval, Fall 2017 Old Dominion University, Computer Science Dept

CS Master's Student Prof: Dr. Nelson

Question 1:

Exercise 4.3:

Try to estimate the number of web pages indexed by two different search engines using the technique described in this chapter. Compare the size estimates from a range of queries and discuss the consistency (or lack of it) of these estimates.

Answer:

I have chosen Google and Bing search engines. The technique described in this chapter uses the formula:

$$N = \frac{f_a \times f_b}{f_{ab}}$$

Where:

a and b are the independent terms

N is the estimated total number of indexed pages

 f_a is the number of pages containing the term a

 f_b is the number of pages containing the term b

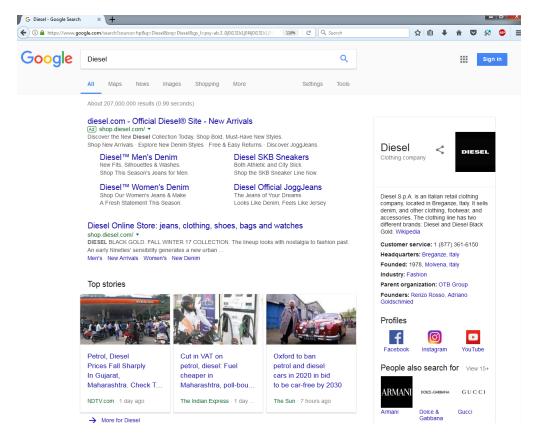
 f_{ab} is the number of pages containing both a and b

The two terms in the query need to be as independent as possible, so I have chosen the following queries:

- 1. Diesel Memory
- 2. Hairy Phone
- 3. Computer Manifold
- 4. Electronic Lumber
- 1. Diesel Memory: I ran the first query "Diesel Memory" and found the following:
 - A. Google:

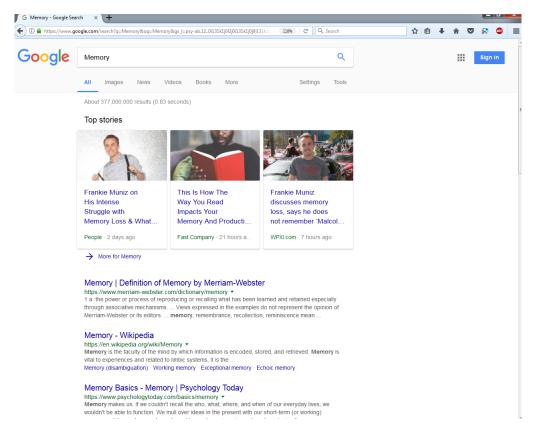
Google found about 207,000,000 results for the term "Diesel".

Figure 1: Query: Diesel, Search Engine: Google



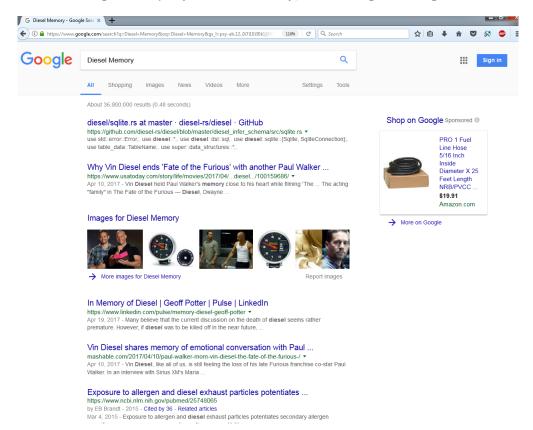
Google found about 377,000,000 results for the term "Memory".

Figure 2: Query: Memory, Search Engine: Google



Google found about 36,900,000 results for the query "Diesel Memory".

Figure 3: Query: Diesel Memory, Search Engine: Google



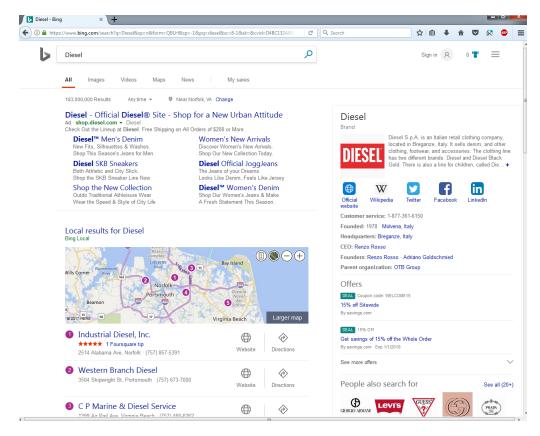
Plugging the results in the formula gives the following:

$$N_{google} = \frac{207,000,000 \times 377,000,000}{36,900,000} = 2114878048.780488 \approx 2114878049$$

B. Bing:

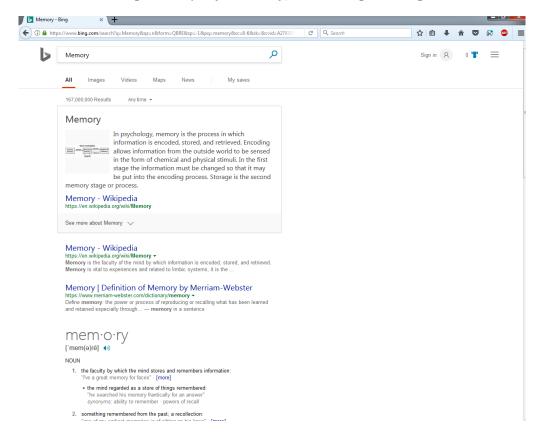
Bing found about 183,000,000 results for the term "Diesel".

Figure 4: Query: Diesel, Search Engine: Bing



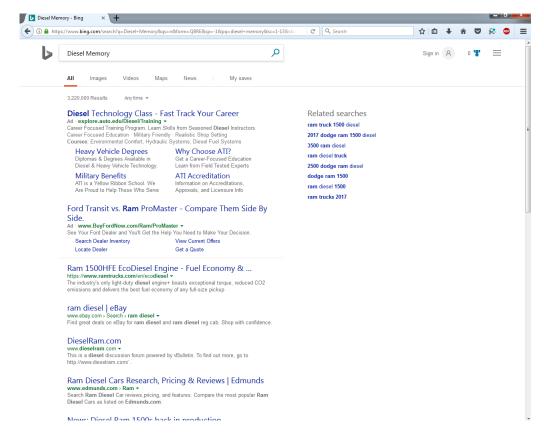
Bing found about 167,000,000 results for the term "Memory".

Figure 5: Query: Memory, Search Engine: Bing



Bing found about 3,220,000 results for the query "Diesel Memory".

Figure 6: Query: Diesel Memory, Search Engine: Bing



Plugging the results in the formula gives the following:

$$N_{Bing} = \frac{183,000,000 \times 167,000,000}{3,220,000} = 9490993788.819876 \approx 9490993789$$

2. Hairy Phone: I ran the second query "Hairy Phone" and found the following: A. Google:

Google found about 106,000,000 results for the term "Hairy".

Google found about 1,410,000,000 results for the term "Phone".

Google found about 8,780,000 results for the query "Hairy Phone".

Plugging the results in the formula gives the following:

$$N_{google} = \frac{106,000,000 \times 1,410,000,000}{8,780,000} = 17022779043.28018 \approx 17022779043$$

B. Bing:

Bing found about 292,000,000 results for the term "Hairy".

Bing found about 748,000,000 results for the term "Phone".

Bing found about 51,500,000 results for the query "Hairy Phone".

Plugging the results in the formula gives the following:

$$N_{Bing} = \frac{292,000,000 \times 748,000,000}{51,500,000} = 4241087378.640777 \approx 4241087379$$

3. Computer Manifold: I ran the third query "Computer Manifold" and found the following:

A. Google:

Google found about 861,000,000 results for the term "Computer".

Google found about 21,500,000 results for the term "Manifold".

Google found about 6,470,000 results for the query "Computer Manifold".

Plugging the results in the formula gives the following:

$$N_{google} = \frac{861,000,000 \times 21,500,000}{6,470,000} = 2861128284.38949 \approx 2861128284$$

B. Bing:

Bing found about 448,000,000 results for the term "Computer".

Bing found about 38,400,000 results for the term "Manifold".

Bing found about 35,100,000 results for the query "Computer Manifold".

Plugging the results in the formula gives the following:

$$N_{Bing} = \frac{448,000,000 \times 38,400,000}{35,100,000} = 490119658.1196581 \approx 490119658$$

4. Electronic Lumber: I ran the forth query "Electronic Lumber" and found the following:

A. Google:

Google found about 1,010,000,000 results for the term "Electronic".

Google found about 43,300,000 results for the term "Lumber".

Google found about 5,720,000 results for the query "Electronic Lumber".

Plugging the results in the formula gives the following:

$$N_{google} = \frac{1,010,000,000 \times 43,300,000}{5,720,000} = 7645629370.629371 \approx 7645629371$$

B. Bing:

Bing found about 289,000,000 results for the term "Electronic".

Bing found about 42,400,000 results for the term "Lumber".

Bing found about 244,000,000 results for the query "Electronic Lumber".

Plugging the results in the formula gives the following:

$$N_{Bing} = \frac{289,000,000 \times 42,400,000}{244,000,000} = 50219672.13114754 \approx 50219672$$

The estimated number of indexed pages in Google as I found are:

2114878049

17022779043

2861128284

7645629371

It is clear that the estimates varied markedly, by few billions, for different queries. I assume that one of the reasons behind that is the difference in "independence level" of the two terms among different queries.

The estimated number of indexed pages in Bing as I found are:

9490993789

4241087379

490119658

50219672

The same inconsistency held for Bing. The estimates varied by few billions.

Question 2:

Exercise 4.6:

Process five Wikipedia documents using the Porter stemmer and the Krovetz stemmer. Compare the number of stems produced and find 10 examples of differences in the stemming that could have an impact on ranking.

Answer:

I have chosen the following documents from WikiSmall collection to process them using Porter and Krovetz stemmers.

Aakrosh_(1998_film).html

Aargau_frank.html

Aaron_Chalmers_3ab5.html

Baba_Budan_67b5.html

Babs_Fafunwa_5f48.html

I wrote a simple python script PKstem.py to process the documents, stem the unique words in them, using Porter and Krovetz stemmers. The stemmers were not implemented from scratch because there are python libraries for both stemmers already.

Listing 1: The content of PKstem.py

```
import sys
import os
import io
import html2text
from nltk import PorterStemmer
import krovetzstemmer
if __name__ == "__main__":
   if len(sys.argv) < 2:</pre>
  print "Usage: python PKstem.py <file1> <file2> <file3> ..."
  print "e.g: python PKstem index.html myPage.html default.html some_page.html"
file_names = set()
for i in range(1,len(sys.argv)):
  file_names.add(sys.argv[i])
porter = PorterStemmer()
krovetz = krovetzstemmer.Stemmer()
output = {}
for file_name in file_names:
   print "Stemming: " + file_name
   h = html2text.HTML2Text()
   h.ignore_links = True
   text = h.handle(u' '.join([line.strip() for line in io.open(file_name, "r",
       encoding="utf-8").readlines()]))
   words = set()
   for word in text.split():
       if word.isalpha():
           words.add(word.lower())
```

```
words = sorted(words)
   porter_output = set()
   krovetz_output = set()
   for word in words:
      porter_output.add(porter.stem(word))
      krovetz_output.add(krovetz.stem(word))
   porter_output = sorted(porter_output)
   krovetz_output = sorted(krovetz_output )
   output[file_name] = {}
   output[file_name]['original_text'] = u' '.join(words)
   output[file_name]['porter'] = u' '.join(porter_output)
   output[file_name]['krovetz'] = u' '.join(krovetz_output)
print "Unique words in the original text:"
print "-----"
print output[file_name]['original_text']
print "Unique stems from Porter stemmer:"
print output[file_name]['porter']
print "Unique stems from Krovetz stemmer:"
print "-----"
print output[file_name]['krovetz']
print "The number of unique words in the original text:"
print "-----"
print len(set(output[file_name]['original_text'].split()))
print "The number of unique stems from Porter stemmer:"
print "-----"
print len(set(output[file_name]['porter'].split()))
print "The number of unique stems from Krovetz stemmer:"
print "-----"
print len(set(output[file_name]['krovetz'].split()))
```

The program accepts the document(s) file name(s) as command line argument(s). It is possible to run the program and process all files at once, however, I ran each file separately.

After running the documents, I found the following results:

Aakrosh_(1998_film).html

Unique words: 229 Porter stems: 224 Krovetz stems: 227 **Aargau_frank.html** Unique words: 199 Porter stems: 193 Krovetz stems: 193

Aaron_Chalmers_3ab5.html

Unique words: 154 Porter stems: 149 Krovetz stems: 150

Baba_Budan_67b5.html

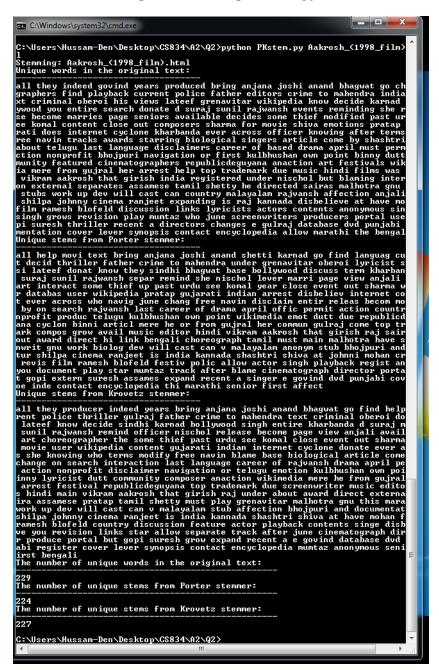
Unique words: 92 Porter stems: 90 Kroyetz stems: 91

Babs_Fafunwa_5f48.html

Unique words: 239 Porter stems: 231 Krovetz stems: 235

This is a screen shot of running the program on the first file:

Figure 7: Running PKstem.py



I found the following examples of differences in the stemming that could have an impact on ranking:

Original Text	Porter Stem	Krovetz Stem
daily	daili	daily
changes	chang	change
academic	academ	academic
collaboration	collabor	collaboration
manchester	manchest	manchester
denominations	denomin	denomination
circulating	circul	circulate
article	articl	article
versus	versu	versus
association	associ	association

Question 3:

Exercise 4.8:

Find the 10 Wikipedia documents with the most inlinks. Show the collection of anchor text for those pages.

Answer:

Question 4:

Exercise 5.8:

Write a program that can build a simple inverted index of a set of text documents. Each inverted list will contain the file names of the documents that contain that word.

Suppose the file A contains the text the quick brown fox, and file B contains the slow blue fox. The output of your program would be:

% ./your-program A B blue B brown A fox A B quick A slow B the A B

Answer:

Question 5:

Exercise 5.10:

Suppose a company develops a new unambiguous lossless compression scheme for 2-bit numbers called SuperShrink. Its developers claim that it will reduce the size of any sequence of 2-bit numbers by at least 1 bit. Prove that the developers are lying. More specifically, prove that either:

SuperShrink never uses less space than an uncompressed encoding, or

There is an input to SuperShrink such that the compressed version is larger than the uncompressed input

You can assume that each 2-bit input number is encoded separately.

Answer:

We have 4 possible 2-bit numbers and 3 possible shorter bit sequences. Let's apply the pigeonhole principle, which states that if n items are put into m containers, with n ξ m, then at least one container must contain more than one item. This means that any mapping from 2-bit sequences to shorter sequences must have at least 2 sequences compressed to the same shorter sequence. Therefore, when trying to decompress this shorter sequence, mentioned above, it is impossible know which of the original 2-bit sequences it was compressed from.

References

- [1] Stackoverflow. https://stackoverflow.com/questions/tagged/python.
- [2] Sergey Brin and Larry Page. Google search engine. http://google.stanford.edu.
- [3] http://www.dba-oracle.com/oracle_news/news_ebay_massive_oracle.htm
- [4] https://www.wired.com/2012/01/amazon-dynamodb/
- [5] https://www.youtube.com/watch?v=a0OvgTfF8Pg
- $[6] \ https://stackoverflow.com/questions/6526181/write-a-program-that-takes-text-as-input-and-produces-a-program-that-reproduces/6526266\#6526266$
- [7] https://en.wikipedia.org/wiki/Pigeonhole_principle