# **Assignment #2**

#### **Homework Submission Guidelines**

- 1. Due date: 26.12.24
- 2. The assignment can be done in pairs
- 3. Answers can be submitted either in English or Hebrew
- 4. HW submission should be done via moodle in the corresponding area (by **only** one of the students)
- 5. Late submission penalty (5% a day) for submitting after the assignment's due date
- 6. Questions / clarifications and more in the dedicated discussion sub-forum in Piazza

## **Dry part (70%)**

## **Vector space model (10%):**

The following matrix represents the word frequencies of four documents d1, d2, d3, d4. Columns represent the documents in the above order; rows represent the vocabulary of six indexed terms a,b,c,d,e,f in that order. (Use **In**.)

	d1	d2	d3	d4
а	0	1	1	1
b	1	2	0	1
С	2	0	0	0
d	0	0	0	0
е	1	0	1	1
f	7	5	7	2

Assume that the fraction of corpus documents in which each term appears is 10%, 10%, 20%, 5%, 50%, 90% for the terms a, b, c, d, e, and f, respectively.

- 1. Compute the cosine similarity between d1 and d2 where terms are represented by the tf-idf scheme. (Describe the tf-idf scheme you have used and provide details of the computation. Use **raw tf**.) (5%)
- 2. Rank the documents in response to the query "a b f". Use the vector space model where document terms are represented by tf and query terms by tf-idf. Provide details of your computations. (Use **raw tf**.) (5%)

# Term Weighting (10%):

- 1. What causes the short-documents bias effect when using cosine similarity? (5%)
- 2. Given following weighted tf function:

$$W_{t,d} = \alpha + (1 - \alpha) \frac{t f_{t,d}}{t f_{\max(d)}}$$

Where  $\alpha$  is a value between 0 and 1;  $tf_{\max(d)}$  is the raw tf of the most frequent term in the document d.

State **one** reason why this weighted tf function is useful and **one** issue that might arise from it. (5%)

# Relevance feedback and evaluation (15%)

1. User 'A' submitted a query to a search engine and obtained an ordered result list. Then, the user provided feedback to the engine (4 – the document is highly relevant to the information need expressed by the query, 0 – the document is not relevant)

DocID	Relevance		
5	4		
2	1		
1	1		
3	3		
4	0		

The total number of relevant documents in the collection is 10. Calculate the **AP**, **precision** and **recall** (at rank 5) (5%)

- 2. Suggest a version of Rocchio's model that utilizes graded relevance judgments. (5%)
- 3. Suggest a version of Rocchio's model that utilizes the rank of relevant documents in the list. (5%)

#### Evaluation (25%)

- Explain the problem of using NDCG for evaluation when negative labels (negative relevance judgments) for documents are used (10%)
- 2. In what cases evaluation using MAP will yield the same results as evaluation using MRR? Mention at least 4 different cases (5%)
- 3. Name two different examples where:
  - a. The removal of stopwords reduces the recall. (2%)
  - b. The removal of stopwords reduces precision. (2%)
- 4. Given a query for which there are 3 relevant documents in the collection.

The R-precision of the search engine for this query is  $\frac{2}{3}$ . Answer the following bullets based on this information.

- 4.1 What is the F1 score at rank 3 for this query? (3%)
- 4.2 What are the maximum and minimum **possible** AP (average precision) values at rank 3 for this query? (You should provide possible values and not general bounds.) (3%)

## True/False questions (10%):

Mark each of the following sentences as true or false and give a short **(but full)** explanation for why your answer is correct:

- 1. p@k is a monotonic non-decreasing function with respect to k. (2%)
- 2. Vector space-based retrieval is always more effective than Boolean retrieval. (1%)
- 3. In the vector space model, the higher the value of the normalization factor for a document is, the lower are the chances of retrieval for that document. (2%)
- 4. The stemming process increases the number of unique terms in the index (2%)
- 5. Values of beta>1 in F-measure emphasize precision. (1%)
- 6. In Rocchio's model,  $q_0$  might be closer to the centroid of the relevant documents than  $q_m$ . (2%)

## Wet part - Intro to Pyserini (30%)

# Part A: (Assignment\_2/data/WET\_PART\_A)

- 1. The documents to be indexed for Part A are located in the file docs.txt
- 2. Create a pyserini index using the following parameters: (change the boldfaced text accordingly)

python -m pyserini.index.lucene --collection TrecCollection --input **The path to the WET\_PART\_A directory** --index **path\_to\_write\_the\_created\_index**--keepStopwords --stemmer none --storePositions --storeDocvectors --storeRaw
--optimize

Do the following to check if the index is created correctly:

## Run the following code:

```
from pyserini.index.lucene import IndexReader
part_a_index = IndexReader('Path to your created index')
print(part_a_index.stats())
```

Desired output:

```
{'total_terms': 212, 'documents': 4, 'non_empty_documents': 4, 'unique_terms': 140}
```

Run retrieval using BM25 method using the following code:

```
# perform retrieval
from pyserini.analysis import Analyzer, get_lucene_analyzer
from pyserini.search import LuceneSearcher
# Initialize the searcher
searcher = LuceneSearcher ('Path to your created index')
# specify custom analyzer for the query processing step to match the
way the index was built
analyzer = get_lucene_analyzer(stemmer='none', stopwords=False) #
Ensure no stopwords are removed from the query
searcher.set_analyzer(analyzer)
# Set BM25 parameters
searcher.set_bm25(kl=0.9, b=0.4)
# perform a search
query = 'corporation'
hits = searcher.search(query, k = 4)

#Display results
for i in range(len(hits)):
    print(f"Doc {hits[i].docid}, Score: {hits[i].score}")
```

- 1. Run a guery "corporation" over the collection using the above parameter file
  - a. How many documents did you retrieve?
  - b. How many documents did you expect to retrieve? Perform and explain the change that is needed for getting the additional documents. (Examine the text of documents.)
- 2. Write a query that will return document D1 first; **use up to 2 words**; explain your choice.
- 3. By running the query: "Michael Jackson" you will retrieve document D4.
  - a. Do you think D4 is relevant to the information need expressed by this query? Explain.
  - Type a query for which D4 can be marked as relevant document; use up to 2 words; explain (refer to the ranking score assigned to D4 in response of the two queries)

#### Part B:

- 1. The files for PartB are located in **Assignment\_2/data/WET\_PART\_B/**
- 2. In the PartB folder you will find the following files and directories:
  - a. "AP\_Coll.tgz" compress file contains AP documents ("database")
    - i. Untar/Unzip the file before indexing
  - b. "queries.txt" query file with 150 queries (stopwords from the query terms were removed in advanced)
  - c. "qrels\_AP" file the AP relevance judgments
  - d. "StopWords.xml" the INQUERY 418 stopwords list

- 3. Build 2 indexes using the given "database" directory
  - a. Index 1: With stopwords removal and with stemming (use "Krovetz" stemmer)
  - b. Index 2: With stopwords removal and without stemming.

**Note:** use the parameter **--stopwords "Path to stopwords.txt file"** to remove the terms in the file from the index.

4. Run retrieval over the 2 created indexes using the following code:

#### Load queries:

```
# Load TSV-format topics
topic_file_path =
r'C:\IDC\Semesters\Winter24\HWs\HW2\WET_PART_B\queries.txt'
topics
=get_topics_with_reader('io.anserini.search.topicreader.TsvIntTopicRea
der',topic_file_path)
#fix query ids
queries = {}
for topic_id, topic in topics.items():
    fixed_topic_id = str(topic_id)
    if len(fixed_topic_id) == 2:
        fixed_topic_id = '0'+str(topic_id)
    queries[fixed_topic_id] = topic['title']
assert len(queries) == 150, 'missing queries'
```

#### For the stemmed index use:

```
# Initialize the searcher with the path to your stemmed index
index path = 'Path to your created index'
searcher = LuceneSearcher(index_path)
# specify custom analyzer for the query processing step to match the
way the index was built
analyzer = get_lucene_analyzer(stemmer='krovetz', stopwords=False) #
Ensure no stopwords are removed from the query
searcher.set_analyzer(analyzer)
# Optionally, configure BM25 parameters (can be adjusted as needed)
searcher.set_bm25(k1=0.9, b=0.4)

#Loop through each query in the topics dictionary and retrieve
documents:
results = {} # To store results for each query
for topic_id, topic in queries.items():
    hits = searcher.search(topic,k=1000) # k=1000 is the number of
retrieved documents
    # Store results in TREC format for each topic
    results[topic_id] = [(hit.docid, i+1, hit.score) for i, hit in
enumerate(hits)]

# Now you can save the results to a file in the TREC format:
output_file = 'Path to write results.items()))
```

```
with open(output_file, 'w') as f:
    for topic_id, hits in sorted_results.items():
        for rank, (docid, _, score) in enumerate(hits, start=1):
            f.write(f"{topic_id} Q0 {docid} {rank} {score:.4f}
pyserini\n")
```

For the un-stemmed index use:

change the index path to the created unstemmed index. In addition change the analyzer to the following:

```
analyzer = get_lucene_analyzer(stemmer='none', stopwords=False) #
Ensure no stopwords are removed from the query
```

5. Use the trec\_eval application or any other evaluation toolkit (details can be found in the lecture of week 2) to evaluate the effectiveness of the 2 retrieved lists and complete the following table. Which retrieval result obtained the highest MAP value? Explain.

Stopword Removal	Krovetz Stemmer	MAP	P@5	P@10
With	With			
38741	18741			
With	Without			

## **Submission Instructions:**

- 1. A **PDF** file containing all answers to the questions (Dry and Wet parts).
- 2. The name of the file as follows: HW2\_Student\_1\_EMAIL\_Student\_2\_EMAIL.pdf