CE706 - Information Retrieval SU 2023

Assignment 1

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Instructions for running your system:

The system is built with the help of Docker Desktop, a platform that allows to build, test and deploy applications quickly. Firstly, I loaded Elasticsearch version 5.6.16 in Docker's container and starting all the actions of the same. We then run our Elasticsearch on jupyter notebook by importing JSON and Elasticsearch libraries (elasticsearch5). After the connection is made, the localhost for Elasticsearch is launched on a web browser to check if it is up and running. We are using python programming language to build a code on further task of the assignment.

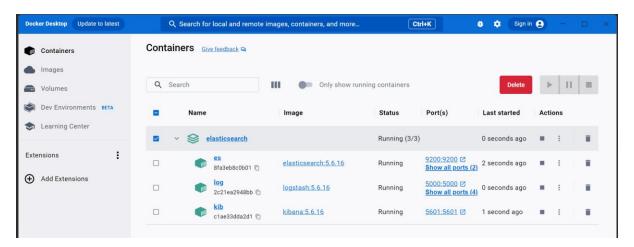


Fig 1.1 Docker Desktop Container with Elasticsearch

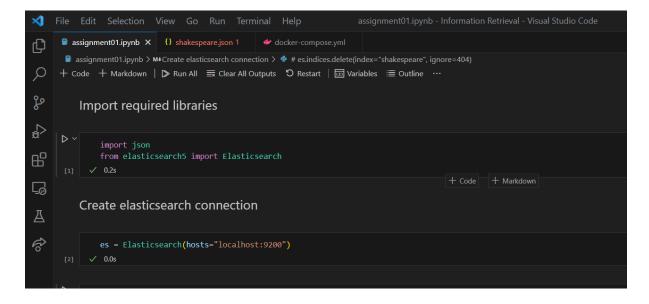


Fig 1.3 Necessary library import and connection with Elasticsearch localhost

Fig 1.2 Elasticsearch server

Indexing:

The dataset is downloaded from Elasticsearch website with version 5.5 which is EOL date.

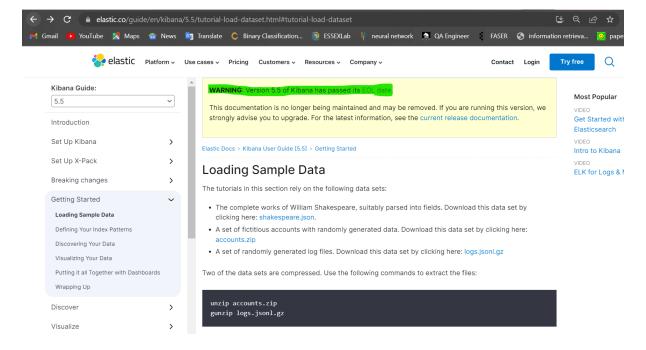


Fig 2.1 Dataset download

The Shakespeare dataset is not available under latest version of Elasticsearch to download.

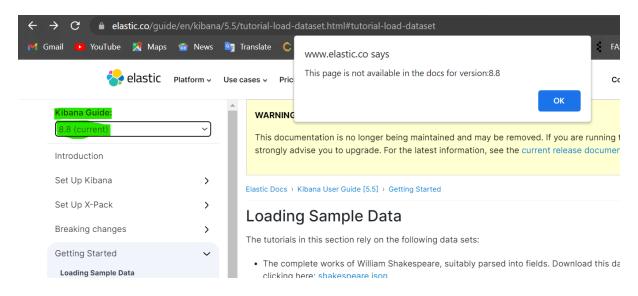


Fig 2.2 Dataset on latest version

The downloaded dataset file is manually extracted and saved in a separate folder created for my Information Retrieval Assignment 01.

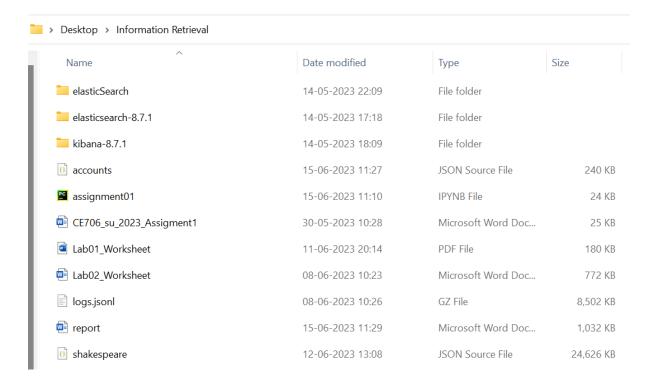


Fig 2.3 Necessary files and dataset for further analysis on Information Retrieval

Our dataset consists of three different fields for '_type': as 'act, 'line', 'scene' and multiple type mapping is unsupported in Elasticsearch from versions 6 and above. Hence, we are using the lower version 5.6.16. Before concluding on to the use of older v5.6.16, several attempts with other versions like v8.8 (current), v7.17, v6.6 were made but it did not result as expected.

Mapping the properties using the default mapping method of Elasticsearch for the dataset. This set of properties are stored in a variable, named "document_mapping". As mentioned earlier, for different fields

of dataset, we are assigning all three fields to another variable and passing on the property's variable into this field's mapping.

```
Document mapping
    document_mappings = {
        "properties": {
            "line_id": {"type": "long"},
             "play_name": {"type": "text'
             "line_number": {"type": "text"},
            "speaker": {"type": "text"},
            "speech number": {"type": "text"},
            "speaker": {"type": "text"},
            "text_entry": {"type": "text"},
    mappings = {
         'mappings": {
            "line": document_mappings,
            "scene": document mappings,
            "act": document mappings,
    mappings
```

Fig 2.3 Document Mapping for the dataset.

Creating Index with the existing functions of Elasticsearch indices and create by passing on the index value, body values that is stored under mapping section and Ignore set to 400 meaning if any error exists, do not fail the execution but continue running the program.

View the created index by making use of the .get_mapping() method for the created document mapping and the JSON.dumps() method that allows to convert a python object into an equivalent JSON object.

Adding Dataset document into index of Elasticsearch. The Shakespeare document is added into the index of created localhost connection with required parameters.

```
Add document to index
    with open("shakespeare.json") as works_of_shakespeare:
        current_line = 1
        for line in works of shakespeare:
            if current line % 2 > 0:
                index info = json.loads(line.strip())
                document = json.loads(line.strip())
                index = index_info["index"]["_index"]
                doc_type = index_info["index"]["_type"]
                doc_id = index_info["index"]["_id"]
                es.index(index=index, doc_type=doc_type, id=doc_id, body=document)
            current line += 1
            if current_line > 300:
                print("done")
                break
    index stats = es.indices.stats(index=shakespeare index)
    index_stats["_all"]["primaries"]["docs"]["count"]
```

Fig 2.4 Adding document to created index.

Tokenization and Normalization

In order to further analyse the data, we are making use of Custom analyser. For Tokenization, standard method is applied and for Normalization the Lowercase token filter is applied to convert all the text entries in the lowercase format. With the analyser, it first performs tokenization on the given number of documents i.e by converting each word of the text entry into tokens and filters it by changing all the mismatched cases of texts into alphabets of lowercase.

After custom analyser is created, a new copy of mapping for the custom analyser is created and the custom analyser is tagged along into this new mapping of document. Settings methods is used to configure the customer analyser into the mapping for all "_type" fields of the document.

Deleting the older index and create a new index with the above configured custom analyser. A test sample for the same has been mentioned in another section.

```
Test analyzer
D
        analyzer_output = es.indices.analyze(
            index=shakespeare_index,
            body={
        tokens = analyzer_output["tokens"]
        for token in tokens:
           print(token["token"])
     and
     breathe
     short
     winded
     accents
     12
     of
     new
     broils
```

Fig 3.1 Example of Tokenization and Normalization (case folding)

For the given above example of standard analyser, the text's output is by converting all the uppercase letters to lowercase and removes all the special characters.

Stemming or Morphological Analysis

For stemming analysis, I'm applying "porter_stem" method that removes common morphological and inflexional endings from the words of English. This method is added into the same mapping of tokenization and case folding but as a next step. An update of the mapping is configured using settings method for all

the fields of "_type" similar to the setting configuration of tokenization and stemming section.

```
    Test analyzer

       analyzer_output = es.indices.analyze(
           index=shakespeare_index,
           body={
               "text": "There are many buses available for route NH77 playing running bathing eating quickly",
               "analyzer": "my_custom_analyzer",
       tokens = analyzer_output["tokens"]
       for token in tokens:
          print(token["token"])
   there
    mani
    buse
    avail
   rout
    nh77
   play
    bath
    eat
    quick
```

Fig 4.1 Stemming/Morphological example

A minor mistake while working with custom analyser "Settings" was made which did not output the stemming on given text entry. i.e., Instead of the keyword "Settings" into the configuration I entered "Setting" which resulted as below.

Fig 4.2 Error on configure setting for stemming

```
Test analyzer
       analyzer_output = es.indices.analyze(
           index=shakespeare_index,
           body={
               "text": "There are many buses available for route NH77 playing running bathing eating quickly",
              "analyzer": "my_custom_analyzer",
       tokens = analyzer_output["tokens"]
       for token in tokens:
          print(token["token"])
   there
    are
   many
   buses
   available
   for
   route
   nh77
   playing
   running
   bathing
   eating
    quickly
```

Fig 4.2 Output of error on configure setting for stemming

Selecting keywords

The selected keywords for our further analysis on dataset are "ngram", "stopword removal" and finding "tf.idf" scores.

In existing custom analyser, I added "stop" keyword into the filter to remove the **stopwords** and the same is configured on the mapping of the document with Settings and a test on the mapped analyser is performed.

Fig 5.1 Selecting keywords - Stopword Removal

```
Test Analyzer for stopword removal

analyzer_output = es.indices.analyze(
    index=shakespeare_index,
    body={"text": "The baby is playing 123!!!!", "analyzer": "my_custom_analyzer"},
)
tokens = analyzer_output["tokens"]
t = [token["token"] for token in tokens]
print(t)

[29]
... ['babi', 'plai', '123']
```

Fig 5.2 Example for stopword removal along with other filters

In the same custom analyser, I'm adding the **N-Gram** filter as "bigram" with minimum and maximum values set to 2 respectively assigned to "type" as "ngram". The updated custom analyser is then configured to mapping of the document using Settings. An example is mentioned below in Fig 5.4.

```
N-Gram

# # add stemming to custom analyzer

# create custom analyzer for n-gram

custom_analyzer = {
    "analysis": {
        "my_custom_analyzer": {
            "type": "custom",
            "tokenizer": "standard",
            "filter": ["lowercase", "stop", "porter_stem", "bigram"],
        }
    },
    "filter": {"bigram": {"type": "ngram", "min_gram": 2, "max_gram": 2}},
    # update elastic search config
    elastic_search_config["settings"] = custom_analyzer
    elastic_search_config
    v   0.1s
```

Fig 5.3 Custom analyser update for N-Grams

```
Test analyzer for ngrams

analyzer_output = es.indices.analyze(
    index=shakespeare_index,
    body={"text": "Children are playing 123!!!!", "analyzer": "my_custom_analyzer"},
)
    tokens = analyzer_output["tokens"]
    t = [token["token"] for token in tokens]
    print(t)

... ['ch', 'hi', 'il', 'ld', 'dr', 're', 'en', 'pl', 'la', 'ai', '12', '23']
```

Fig 5.4 Example for N-Gram tokenizer

In the above example, the text line is first performs standard tokenization, converts them to lowercase, removes all the special characters present in the line, performs stemming, removes English stopwords, and outputs a list of words in list of bigrams.

TD.IDF

After all the custom analysers are completed and executed successfully, we are then trying to configure tf.idf score using Elasticsearch similarity module. The term frequency of how many times a given word appears in the document and inverse documentary frequency calculation is the score that how rare your word is in the corpus. This score is obtained by the in-built functionality of Elasticsearch called Classic Similarity. I have stored this similarity methods as a dictionary to a variable and then passing on this dictionary variable to the index of mapping with different fields of text_entry.

Fig 5.5 Similarity module TF.IDF

Note: For every time the custom analyser is updated with new token or filter i.e, updating custom analyser by adding porter_stem, stop, lowercase, I'm deleting the created indices and creating a new index each time as Update option to the index is not available.

Once all the task of pre-processing steps has been completed, I'm importing the Shakespeare dataset file into the index with 1000 lines in order to perform search queries. This step is basically Indexing 1000 documents from the Shakespeare dataset.

Searching

The Elasticsearch search queries used are search on full text, search with exact phrase, match exact phrase using AND operator on particular fields, search query to match part of phrase (prefix), sorting the field "line_id" in descending order and additionally Pagination and Filter part of phrase is also executed.

Fig 6.1 Search on full text for query "pagans" is resulted

```
Search query to match exact phrases

***Ext_entry*: ("query*: "twy love"))))

resp = es.scarch(index="shakespeare", body-match_phrase)

for hit in resp["hits"]! "hits"]:

print(hit)

***Print(hit)

***
```

Fig 6.2 Search query to match exact phrase "thy love" on text_entry field is executed

Fig 6.3 Search query to match phrases on multiple fields using AND operator

Fig 6.4 Search query to match part of phrase on multiple fields