Project Description: Information Retrieval in a Medical Blog

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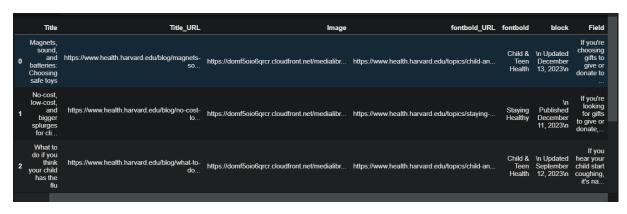
Introduction

Our application consists of a web medical blog. You can create; read and most importantly search about the relevant posts that will reply your query well. The marking feature of the application is the semantic search of the posts.

Methodology

Article scrapping

We scraped posts using Selenium.webdriver from https://www.health.harvard.edu/blog



Encode medical articles using S-bert

Tokenization:

Tokenize the medical articles to convert them into a format suitable for input to the S-BERT model.

Sentence Embedding:

Use the S-BERT model to obtain embeddings for each sentence in the medical articles. S-BERT captures semantic similarity by mapping sentences into a high-dimensional vector space.

```
df.description_vector[:5]

0   [-0.0070151784, 0.0033397812, -0.018809972, 0....
1   [0.041044563, 0.05344586, 0.017541107, 0.07421...
2   [-0.01417845, 0.00091768993, -0.0012399706, 0....
3   [-0.0047708116, 0.0065720384, 0.013290583, -0....
4   [0.025648113, 0.044721454, 0.010994425, -0.055...
Name: description_vector, dtype: object
```

Indexation automatic using elasticsearch

1. Install and Run Elasticsearch in docker:

```
elasticsearch
7f03b6ec0c6f 8.11.0 In use 1 month ago 1.41 GB  

Administrator: Windows Powx × +  

Windows PowerShell
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Install the latest PowerShell for new features and improvements! https://aka.ms/PSWindows

PS C:\Users\Honor> docker run --rm --name elasticsearch_container -p 9200:9200 -p 9300:9300 -e "discovery.type=single-node" -e "xpack.security.enabled=false" elasticsearch:8.11.0
```

2. Set Up an Index Mapping:

```
indexMapping = {
    "properties" : {
        "type" : "long"
    },
    "medical_abstract" : {
        "type" : "text"
    },
    "description_vector" : {
        "type" : "dense_vector",
        "dims" : 768,
        "index" : True,
        "similarity" : "12_norm" # dist euclid
    },
},
}
```

3. Index the Document:

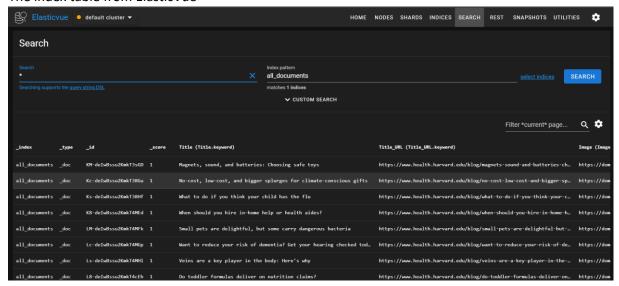
```
Create new index
In [26]: from indexMapping import indexMapping
In [27]: es.indices.create(index="all_documents", mappings= indexMapping)

In [32]: records_list = df.to_dict("records")

In [33]: for record in records_list:
    try:
        es.index(index="all_documents", document=record)
    except Exception as e:
        print(e)

In [34]: es.count(index="all_documents")
Out[34]: ObjectApiResponse({'count': 180, '_shards': {'total': 1, 'successful': 1, 'skipped': 0, 'failed': 0}})
```

The index table from ElasticVue



4. Querying the Index:

```
query = {
    "field": "description_vector",
    "query_vector": vector_input,
    "k":5,
    "num_candidates": 60,
}
```

5. Matching the query with document

```
res = es.knn_search(index="all_documents", knn=query, source=["Title","Field", "Image"])
results = res["hits"]["hits"]

for result in results:
    if "_source" in result:
        try:
        print(f"Document score: {result['_score']}")
        print(f"Document Title: {result['_source']['Title']}")
        print(50*"_")
    except Exception as e:
        print(e)
```

To Match the relevant document with the query we used the euclidien distance in the similarity field of our Mapping in elasticsearch, avec le classement des resultat selon cette distance

UI and Backend part

We used Spring Boot for the backend of our website and ReactJS for the frontend.

Result (using streamlit to test the model)



Future Work

We will parse the index from MongoDB to ElasticSearch using logStach

Integrate the search in the image too by creating a description from the image then index this description and make it available for searching