CSC 8712

A framework for building inverted index on textual data

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Introduction:

Background:

- Data is growing exponentially. There is a need for enhancements in techniques for processing and querying
 of textual data.
- Searching and querying of records are the two bottleneck.

Evolution of searching and querying from large corpus of data:

- 1. RAID (Redundant Array of Independent Disks).
- 2. Hashing(Equality and Membership Queries).
- 3. Indexing(secondary access paths).

Introduction:

What is an Inverted Index:

- An inverted index is the mapping of words to documents. A storage of a list of words/literals and the documents in which they appear.
- An inverted index can also store the occurrence(frequencies) of words or weights.

Inverted Index vs Database Index:

- A list of documents(block of records) and the words that it contains.
- Table of Contents(TOC) vs Sections(End of a book).

Inverted Index vs Database Index

body text 66, 68, 79-82, 92-94, 115, 233-235

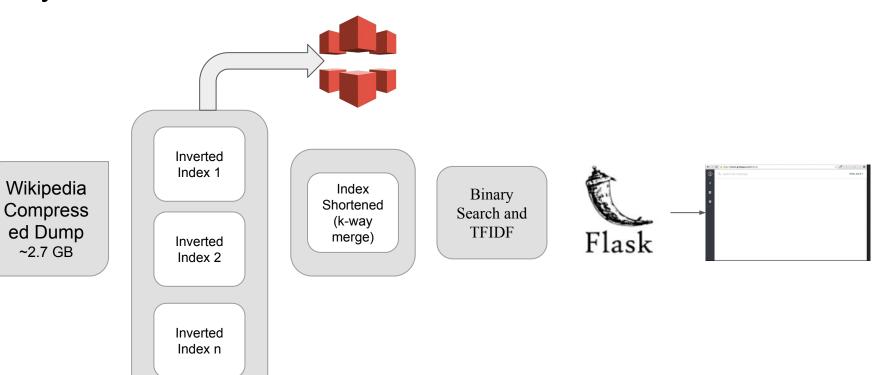
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System Overview:



Methodology:

I. Indexing

- We pre-process the whole data by removing stopwords and applying stemming.
- We create an inverse map of document id to wiki page title.
- The key will be a word and the value will be a list of file_ids (wikipage_id) and how many times this word occurred in that file in title, body, infobox, references etc.
 - apple: {"file_id t2 b3 r4", "file_id1 t2 b3 r4", "file_id2 t2 b3 r4"}.
 - banana : {"file id t4 b6 r3"}.
- As this inverted index can grow very large, we try to do it in a batch wise manner and create an index file for this chunk.

Methodology:

II. Merging Indices:

- We merge these sorted index files into a single file.
- We use K-way merge with min-heap to merge as we cannot load all the data into memory at once.
- Min-heap will help us in merging the sorted files in a reduced complexity of NlogK as the least value is always at the root node.

Main Idea:

Min-heap will return the smallest element from a collection of k-smallest elements from k-sorted lists.

Accordingly, we merge the sorted files in our case in a time complexity of O(NlogK).

Methodology:

III. Querying and Ranking:

- Given a sequence of words, we need to filter the documents, and ranking is done on these documents.
- Field queries such as "Title: Mahatma Body: India" are also supported.
- Based on the filtered documents, we score each document with a cumulative sum of TF-IDF for each word.
- For each word in the query, we search the index file by doing a binary search.

Main Idea: A binary search will help in finding the relevant querying the ranked(sorted) elements in O(log₂N).

Tech Stack:

XML, Python, Flask, HTML, CSS, Amazon S3.

- For purpose of searching, in most cases we use engines like Solr/elasticsearch.
- Indexing also involves implementing some indexing technique. Here we will build index and search from scratch.

Dataset:

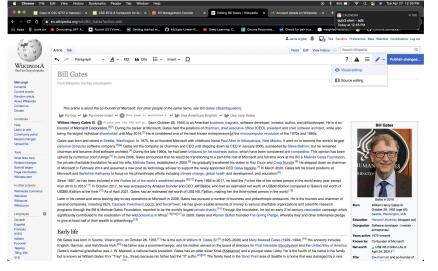
- Wikipedia XML Dump.
 - https://meta.wikimedia.org/wiki/Data_dump_torrents#English_Wikipedia
 - \circ ~2.7GB(zipped)
- Ever growing dataset.
 - Wikipedia ~2500 articles/day, growing at terabytes rate.
- Various Formats: JSON,XML, RDF.
- Open-Source Dataset.
- Articles are separated by a pair of tags as in <<u>page>contents</page></u>. The title of each article is marked by <u>"<title>Name of the article</title>"</u>
 - Links to other Wikipedia articles are of the form "[[Name of other article]]".

Implementation: Data Pre-processing

Bill Gates

children

From Wikipedia, the free encyclopedia ((Short description American business magnate and philanthropist)) {{about | the co-founder of Microsoft | other people of the same name | Bill Gates (disambiguation)}} {{pp-move-indef}} <!-- This page is monitored regularly for nonsense and vandalism. People who vandalize biographies about living people will be blocked from editing. If you would like to experiment with Wikipedia, please copy the following address into your browser's address bar: https://en.wikipedia.org/wiki/ and it will take you to a page where new users can try out the editing ({Use American English|date=March 2020}) {{Use mdy dates|date=April 2021}} {{Infobox_person nane = < !-- Please see the FAQ and previous discussions on the article's talk page. Including post-nominals at the top of the infohox would give them an undue prominence post-nominals in an article about an American who has no close association with the British realm. --> = Bill Gates 2018.jpg<|-- Consensus in talk page for this image. If you want to change it go to talk page --> - Head and shoulders photo of Bill Gates caption = Gates in 2018 birth name - William Henry Gates III = ((birth date and age | 1955 | 10 | 28)) birth date = [[Seattle]], [[Washington (state) | Washington]], U.S. birth place = [[Harvard University]] (dropped out) education alma_mater = <|--If there is anything that questions status of a subject as an alumnus of a university, it is not included in the alma mater/education section (i.e. reference Natalie Portman vs . Mark Zuckerberg)-both attended Harvard, one graduated, the other did not .--> known for = Co-founder of [[Microsoft]] occupation = {{hlist|Software developer|investor|entrepreneur}} = {{\text{UB}}} US\$146.28\text{nhmp:hillion} ({{\text{As of }|2021|04|24|df=US}})<ref name="Bloomberg">{{\text{cite web}|url=https://www.bloomberg.com/billionaires/profiles/william-h-gates/}} net worth |title=Bloomberg Billionaires Index: Bill Gates |publisher=Bloomberg |access-date=April 24, 2021}}</ref> = 1975-present years active title = {{indented plainlist * Co-[[chairman]] and co-founder of the [[Bill & Melinda Gates Foundation]] * Chairman and founder of [[Branded Entertainment Network]] * Chairman and founder of [[Cascade Investment]] * Chairman and co-founder of [[TerraPower]] * [[Technical advisor Technology advisor]] of Microsoft)) boards spouse = {{marriage|[[Melinda Gates|Melinda French]]|January 1, 1994}}



Execution Flow

Load Sample Dataset index.py merge.py compress.py boto_aws_s3.py file_sizes.py search.py

```
2
     import nltk
 3
     nltk.download('stopwords')
     from nltk.corpus import stopwords
     from datetime import datetime
 7
     import os
     import pickle
     import json
10
     import re
11
12
     from nltk import stem
13
     from collections import Counter
     from collections import defaultdict
15
     import xml.sax as sax
16
17
     LEN_INFOBOX = len('{{infobox')}
     LEN_CATEGORY = len('[[category:')
19
     LEN EXTERNS = len('==external links==')
20
     TAG_RE = re.compile(r'<[^>]+>')
21
22
      infoBoxRegex = re.compile(r'(\{\{infobox(.|\n)*?\}\n)(?:[^\|])')
      categoryRegex = re.compile(r'\[\[category:.*\]\]\n')
23
24
      externalsRegex = re.compile(r'==external links==\n(.|\n)*?\n\n')
     reftagRegex = re.compile(r'<ref(.|\n)*?</ref>')
      refsRegex = re.compile(r'(==references==(.|\n)*?\n)(==|\{\{DEFAULTSORT})')
26
     stop words = set(stopwords.words('english'))
27
     stemmer = stem.PorterStemmer()
28
29
     MIN_STOP_LENGTH = 3
     MAX STOP LENGTH = 15
30
31
     alphabet_lis = 'abcdefghijklmnopgrstuvwxyz'
32
33
     # punctuation='i@!"#$%&\'()*+,-./:;<=>?@[\\]^ `{|}~0123456789'
     dert = [ele for ele in list(range(0,1_1141_111)) if ele not in list(range(97,123))]
```

index.py > ...

1

import xml

index.py

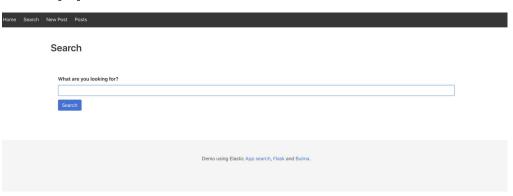
```
parser = xml.sax.make_parser()
handler = WikiHandler()
hellp = parser.setContentHandler(handler)
now = datetime.now()
paths = ['/Applications/Final/data/shortest.xml','/Applications/Final/data/Infosys.xml','/Applications/Final/data/Georgia_S
for path in paths:
    parser.parse(path)
diff = datetime.now() - now
print(diff.total_seconds())
```

merge.py

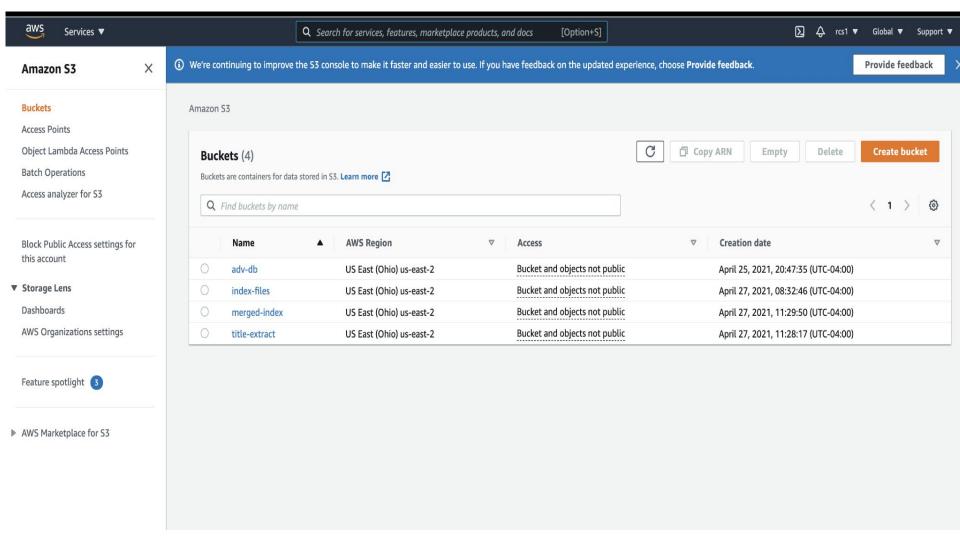
```
def merge(input_files):
   heap = []
   data = ''
   _output_file = open('./index_m','w+')
   if True:
       open_files = []
        [open_files.append(open(os.path.join('./buggi_dir/',file__), 'r')) for file__ in input_files]
       # enqueue the pair (key, sent, f) using the first value as priority key
        for file__ in open_files:
            heapq.heappush(_heap,tuplify(file__))
       while( heap):
            smallest = heapq.heappop(_heap)
            data += smallest[1]
            _output_file.write(smallest[1])
           # read next line from current file
            tuple__ = tuplify(smallest[2])
            # check that this file has not ended
            if(len(tuple__[1]) != 0):
                heapq.heappush(_heap, tuple__)
```

```
output = []
     while True:
         query = input()
         if query == '!':
             break
33
         output = []
         now = datetime.now()
34
35
         if ':' in query: # tital: Mahatma
             output = FieldSearch(query)
36
         else:
             output = NormalSearch(query)
38
39
         diff = datetime.now() - now
90
         print('d', output)
         for ele in output:
92
             print(ele.strip())
         print(diff.total_seconds())
93
```

Application View





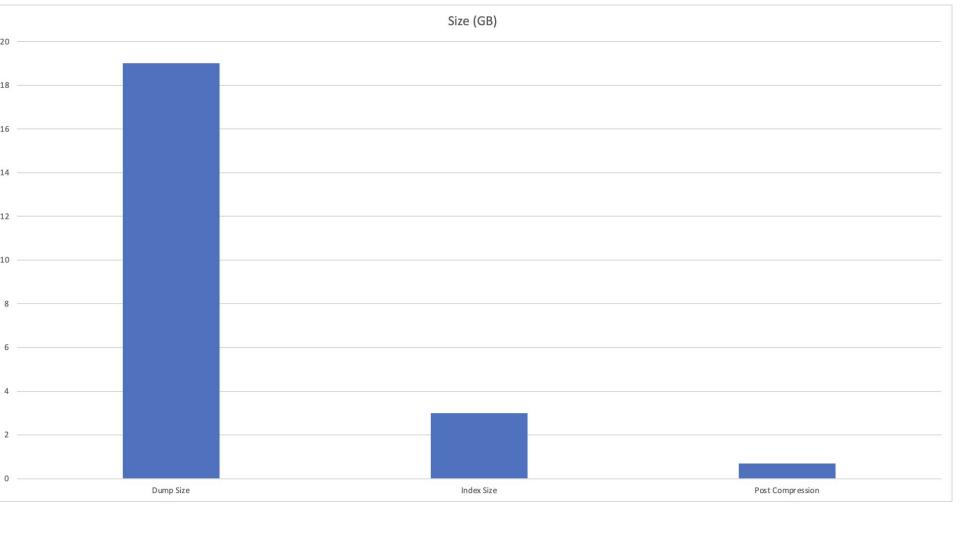


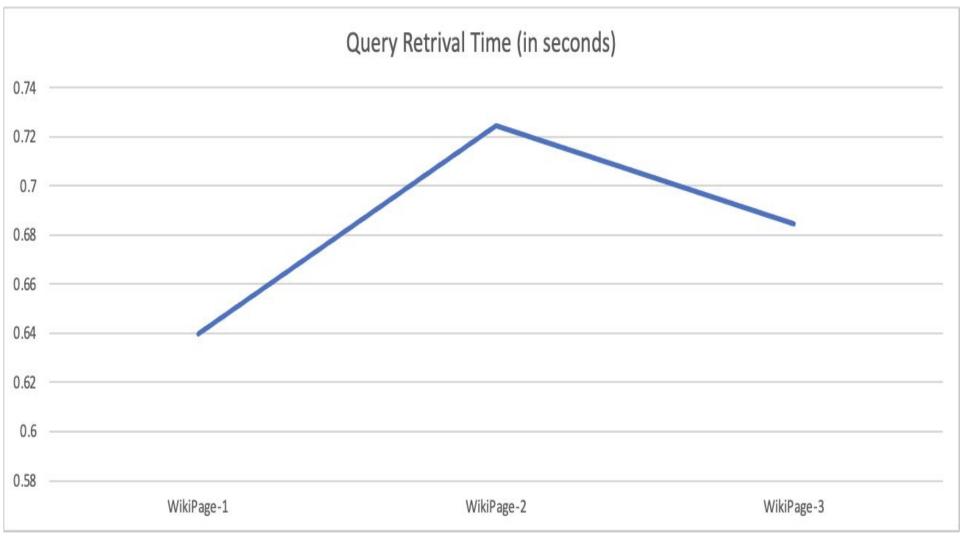
Why this Tech Stack

- Python Most widely used programming language.
- Flask lightweight
- S3 Faster, Cheaper and easier to query
- ElasticSearch preconfigured search cluster.

Evaluation plan

- Developing an application/system to showcase the search performance of a dataset over an inverted index
 - Search performance is the query retrieval time.
- Measuring the time taken to index large textual corpus. (Index creation time)
 - Overall time taken is 1.5 hours for complete dataset
- Inverted index quality: search result quality
 - Correct outputs
- Inverted Index size to be proportionally less than dump size.
 - Graph in next slide
- Scalability of the index
 - The index creation time for any number of articles is constant.





DEMO

Contributions

Gaurav - Merging and Compression, Front End

Moulika - Data Preprocessing, S3 configuration

Rajath - Index Creation and Searching, Flask Backend linking

Future Scope:

Wikipedia History Page is a use-case of a version control system. It stores all the modification(versions) of a wikipedia page.

Creating a reliable inverted index for the history page is a difficult task due to the following reasons:

- Deleted content can have complications on the ranking of the wikipedia pages.
- Copyright infringement issues for a previous versions.

Conclusion:

- A framework for inverted index is useful for fielded and full-text search.
- Use of k-way merge provides an efficient merging sorted inverted index files.
- Perfect use case for a binary search.

Drawbacks:

Ranking of wikipedia pages can depend on various other parameters and the list is exhaustive.

Eg: PageRank algorithm ranks documents based on page links back to a webpage. An assumption is made that an arbitrary webpage is more relevant when it has more links pointing.

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Thank You.

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