

Faculty of Engineering and Technology



Mini Project Report

on

"Twitter Data Sentiment Analysis on Fungus Live Dataset"

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ABSTRACT

In today 's highly developed world, every minute, people around the globe express

themselves via various platforms on the Web. And in each minute, a huge amount of unstructured

data is generated. Such data is termed as big data. Twitter, one of the largest social media site

receives millions of tweets every day on variety of important issues. This huge amount of raw data

can be used for industrial, social, economic, government policies or business purpose by

organizing according to our requirement and processing. Hadoop is one of the best tool options for

twitter data analysis as it works for distributed big data, streaming data, time stamped data, text

data etc. Hence, Flume is used to extract real time twitter data into HDFS. Hive and Pig which is

SQL like query language is used for some extraction and analysis. People's psychic and emotional

wellbeing has been strongly proportional to this pandemic and they are suffering from panic, terror,

and anxiety as the number of cases are increasing at an alarming rate around the world, we thus

have retrieved data from related to the Yellow Fungus, Black & White Fungus.

These tweets have proven to be a valuable source of information in the recent years, playing key

roles in success of brands, businesses and politicians. We have tackled Sentiment Analysis with a

lexicon-based approach for extracting positive, negative, and neutral tweets by using part-of-

speech tagging from natural language processing. We begin by collecting datasets for analysis,

and then, depending on the method, we apply cleaning to it. After cleaning the dataset, we convert

the rows into tokenized words and reflect the important words with the help of stemming &

lemmatizations. The approach manifests in the design of a software toolkit that facilitates the

sentiment analysis.

KEYWORDS: Healthcare · Hadoop · MapReduce · HDFS · Tableau · Big data · COVID-19 ·

Corona virus

Twitter Data Sentiment Analysis on Fungus Live Dataset

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CHAPTER 1 INTRODUCTION

1.1. Introduction

Blogging and networking platforms like Facebook, Reddit, Twitter and LinkedIn are social media channels where users can share their thoughts and opinions. Since online chatter is a vital and exhaustive source of information, these thoughts and opinions hold the key to the success of any endeavour. Tweets which are posted by millions all over the world can be used to analyse consumers "opinions about individual products, services and campaigns.

Lately, people's psychic and emotional wellbeing has been strongly proportional to this pandemic and they are suffering from panic, terror, and anxiety as the number of cases are increasing at an alarming rate around the world. When the second wave of covid emerged this year, it was notable not just because of the disruption it caused, but rather because it occurred suddenly at a time when people were returning to normal. Information of this kind assists public health authorities in identifying popular issues and get a glimpse of all the global health messages sensed in the globe across Twitter. All of these will aid in improved policy design while keeping social perspectives in mind, as well as the government's awareness of issues such as food poverty, vaccination shortages, and so on.

While the country is still dealing with the coronavirus pandemic, cases of black fungus or mucormycosis are on the rise in the country. The cases of white fungus and yellow fungus have also been reported from some parts of the country. On one hand, due to the suddenness and novelty of these fungus, there was a serious lack of knowledge of how deadly this can bw, and it is difficult to quickly develop a cure for it. On the other hand, governments have not taken timely preventive measures to suppress the spread of this, resulting in severe damage to human health and societal stability. In addition, before and after the outbreak, because of the lack of an effective early warning, rapid response mechanisms, implementation of effective prevention and control

decisions, the best prevention time have been missed. In order to respond efficiently and propose a preventative control plan, not only do we need a complete emergency management system, but also scientific data analysis for decision support.

However, with the spread of these diseases, the analysis results based on the gradually generated large amounts of data have played an important role in the tracking of people's movements, early warning of high-risk areas, screening of asymptomatic potential infections, drug development, information release, and policy support. They have also become an important basis for the implementation of preventive control programs and have played an important role in enhancing the modernization level of national governance, promoting protection and improving people's livelihood.

Black Fungus is a term given to the disease called Mucormycosis. It is a fungal infection caused by the Mucorales order in which the species most commonly implicated are Rhizopus, Mucor and Absidia. Mycosis is a term for fungal infection and hence Mucormycosis-Mycosis caused by Mucor.

White fungus is a genus of Yeast called Candida. It grows in the lab as white/ creamy white spots on plates of agar. In humans, they again appear as white, creamy spots on the mucosa of the oral cavity most commonly. They are commensals in the oral cavity and gastrointestinal tract and are present on skin as well. Candida albicans is the most commonly isolated species. Albicans means white. Hence the term white fungus.

Yellow fungus is a term coined by someone and unfortunately does not have any significance. Yellow is the colour of pus, which is formed whenever there is any bacterial infection. Superadded on that, there may be fungal growth of a wide variety of species. It does not mean that the fungal infection is causing the colour. It also most definitely does not mean that there is a "new yellow fungus which is more dangerous than black and white fungus"

These infections have been prevalent before, and we have been dealing with fewer cases, and in select groups having low immunity. These are not new diseases. The sheer numbers though are

significant due to the number of patients with lowered immunity and additional risk factors. There may soon be a "Green fungus" or "Multicoloured fungus" variant reports coming in from different parts of the country.

While the country is still dealing with the coronavirus pandemic, cases of black fungus or mucormycosis are on the rise in the country. Meanwhile, cases of white fungus and yellow fungus have also been reported from some parts of the country.

Moreover, there is still a lack of a theoretical framework for big data analytics in the prevention and control of Major Public Health Incidents (including these fungus). Thus, it is necessary to propose such a framework to focus on the prevention and control of it using big data, which is also applicative in instances of other epidemic diseases. The proposed definitions, characteristics, data sources, applications, and framework can also enlighten and support the decisions of governments, enterprise, medical institutions, users, and researchers.

1.2. Motivation

Little work has been done to actually expand on the topic of the correlation between Twitter sentiment and these new diseases. Today we are living in the world which is surrounded by 99% of data. There are different microblogging sites where users express their views about different products these views are nothing but opinions of people and it will go waste if it is not used in proper way so there is a need to use opinions of people in improving productivity, usefulness, functionality of particular product or application or technique or any entertainment resource. Thus, it is a good idea to use Big Data technologies to perform sentiment analysis.

1.3. Problem definition

The project focuses on using Twitter, the most popular micro blogging platform, for the task of sentiment analysis. The tweets are important for analysis because data arrive at a high frequency and algorithms that process them must do so under very strict constraints of storage and time. It will be shown how to automatically collect a corpus for sentiment analysis and opinion mining purposes and then perform linguistic analysis of the collected corpus. All public tweets posted on twitter are freely available through a set of APIs provided by Twitter. Using the corpus, a sentiment classifier, is constructed that is able to determine positive, negative and neutral sentiments.

1.4. Objectives

Twitter has over a billion users and everyday people generate billions of tweets over 100 hours per minute and this number is ever increasing. To analyse and understand the activity occurring on such a massive scale, a relational SQL database is not enough. Such kind of data is well suited to a massively parallel and distributed system like Hadoop. The main objective of this project is to focus on how data generated from Twitter can be mined and utilized by different companies to make targeted, real time and informed decisions about their product that can decrease the disease or to find out the views of people on a specific topic of interest. This can be done by using Hadoop concepts. The given project will focus on how data generated from Twitter can be mined and utilized. There are multiple applications of this project. Companies, Doctors, Politicians, Decision Makers, etc can use this project to understand how effective and penetrative their plans & programs are through sentiment analysis.

CHAPTER 2 LITERATURE SURVEY

2.1. Existing and proposed system

The major issues involved in big data are the following:

- The first challenge faced is storing and accessing the information from the large huge amount of data sets from the clusters. We need a standard computing platform to manage large data since the data is growing, and data stores in different data storage locations in a centralized system, which will scale down the huge data into sizable data for computing.
- The second challenge is retrieving the data from the large social media data sets. In the scenarios where the data is growing daily, it's somewhat difficult to accessing the data from the large networks if we want to do specific action to be performed.
- The third challenge concentrates on the algorithm design for handling the problems raised by the huge data volume and the dynamic data characteristics.
- The main scope of the project is to fetching and analysing the tweets on Types of fungal disease and to perform sentiment analysis to find the most popular tweets which are trending and finding the sentiment rating of each tweet based on that topic. Sentiment Analysis is the process of detecting the contextual polarity of text. A common use case for this technology is to discover how people feel about a particular topic.

2.2. Background study

These days internet is being widely used than it was used a few years back. Billions of people are using social media and social networking every day all across the globe. Such a huge number of people generate a flood of data which have become quite complex to manage. Considering this enormous data, a term has been coined to represent it. This term is called Big Data. Big Data is the term coined to refer this huge amount of data. The concept of big data is fast spreading its arms all over the world.

2.2.1. Big Data

Data which is very large in size and yet growing exponentially with time is called as big data. It refers to the large volume of data which may be structured or unstructured and which make use of certain new technologies and techniques to handle it

Hadoop is a programming framework used to support the processing of large data sets in a distributed computing environment. It provides storage for a large volume of data along with advanced processing power. It also gives the ability to handle multiple tasks and jobs. Hadoop was developed by Google's MapReduce that is a software framework where an application break down into various parts. The Apache Hadoop ecosystem consists of the Hadoop Kernel, MapReduce, HDFS and numbers of various components like Apache Flume, Apache Hive and Apache Pig which are being used in this project.



Figure 1: Big Data

2.2.1.1. Categories of Big Data

Structured Data: The data which can be stored and processed in table (rows and column) format is called as a structured data. Structured data is relatively simple to enter, store and analyze. Example - Relational database management system.

Unstructured Data: The data with unknown form or structure is called as unstructured data. They are difficult for nontechnical users and data analysts to understand and process. Example - Text files, images, videos, email, webpages, PDF files, PPT, social media data etc.

Semi-structured Data: Semi-structured data is data that is neither raw data nor organized in a rational model like a table. XML and JSON documents are semi structured documents.

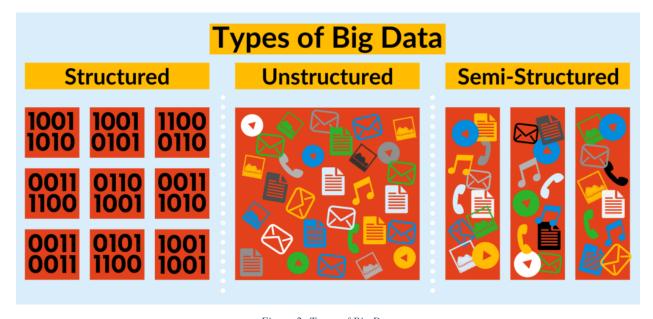


Figure 2: Types of Big Data

2.2.1.2. Characteristics of Big Data

The characteristics of Big Data are defined by three V's:

- **Volume** It refers to the amount of data that is generated. The data can be low density, high volume, structured/unstructured or data with unknown value. The data can range from terabytes to petabytes.
- **Velocity** It refers to the rate at which the data is generated. The data is received at an unprecedented speed and is acted upon in a timely manner.
- **Variety** Variety refers to different formats of data. It may be structured, unstructured or semi- structured. The data can be audio, video, text or email.

2.2.2. Hadoop

As organizations are getting flooded with massive amount of raw data, the challenge here is that traditional tools are poorly equipped to deal with the scale and complexity of such kind of data. That's where Hadoop comes in. Hadoop is well suited to meet many Big Data challenges, especially with high volumes of data and data with a variety of structures.

Hadoop is a framework for storing data on large clusters of commodity hardware, everyday computer hardware that is affordable and easily available and running applications against that data. A cluster is a group of interconnected computers (known as nodes) that can work together on the same problem. The Current Apache Hadoop ecosystem consists of the Hadoop Kernel, Map-Reduce, HDFS and numbers of various components like Apache Hive, Pig, Flume etc.

Hadoop consists of two main components:

- HDFS (Data Storage)
- Map-Reduce (Analysing and Processing)

2.2.2.1. Architecture of Hadoop

HDFS is the main component of Hadoop architecture. It stands for Hadoop Distributed File Systems. It is used to store a large amount of data and multiple machines are used for this storage. MapReduce is another component of big data architecture. The data is processed here in a distributed manner across multiple machines. So, HDFS works as a storage part and MapReduce works as a processing part. Hive and Pig are the components of Hadoop ecosystem. These are high level data flow languages. MapReduce is the inner most layer of Hadoop ecosystem.

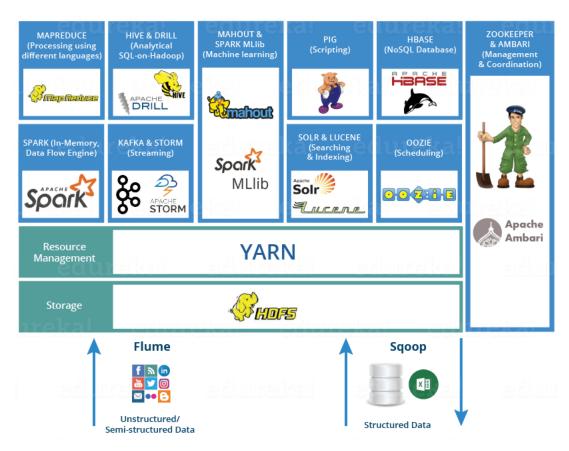


Figure 3: Architecture of Hadoop

2.2.3. Technologies Used

Apache Flume: Apache Flume is a distributed, reliable, and available service for efficiently collecting, aggregating, and moving large amounts of streaming data into the Hadoop Distributed File System (HDFS). It can be used for dumping twitter data in Hadoop HDFS. It has a simple and flexible architecture based on streaming data flows; and is robust and fault tolerant with tuneable reliability mechanisms for failover and recovery. Flume lets Hadoop users ingest high-volume streaming data into HDFS for storage.

Apache Hive: Hive is a data warehouse infrastructure tool to process structured data in Hadoop. It resides on top of Hadoop to summarize Big Data, and makes querying and analysing easy. Hive provides the ability to store large amounts of data in HDFS. Hive was designed to appeal to a community comfortable with SQL. Hive uses an SQL like language known as HIVEQL. Its philosophy is that we don't need yet another scripting language. Hive supports maps and reduced transform scripts in the language of the user's choice which can be embedded with SQL. Supporting SQL syntax also means that it is possible to integrate with existing tools like. Hive has an ODBC (Open Database Connectivity JDBC (Java Database Connectivity) driver that allows and facilitates easy queries. It also adds support for indexes which allows support for queries common in such environment. Hive is a framework for performing analytical queries. Big Data enterprises require fast analysis of data collected over a period of time. Hive is an excellent tool for analytical querying of historical data. It is to be noted that the data needs to be well organized, which would allow Hive to fully unleash its processing and analytical powers.

When you are looking to process clusters of unorganized, unstructured, decentralized data and don't want to deviate too much from your solid SQL foundation, Pig is the option to go with. You no longer need to get into writing core MapReduce jobs. If you already have SQL background, the learning curve will be smooth and development time will be faster.

2.3. Dataset Description

In this proposed method, Live Dataset is retrieved from twitter using flume related to the keywords: black fungus, white fungus, yellow fungus, fungus, fungai, Mucormycosis, Mucor, zygomycosis, mucoromycetes, Candidiasis, Candida, aspergillosis. The dataset comprises of factual information, results of examinations and information given by the users. The coloumns which are present in the dataset are Tweets, User, user statuses count, user followers, user location, user verified, fav count, retweet count, tweet date, sentiment. Some of the subjective features like fav count and retweet count binary outputs, user status count which is an examination feature showcases three different scenarios wherein helping in understanding user's activity. Tweets reflects the chances of understanding what are people's sentiment and ideation over fungal infection.

The dataset is divided into smaller blocks which are primarily processed by "Map Phase" in parallel and then by "Reduce Phase". Hadoop framework has sorted out the output of the Map phase which are then given as an input to Reduce Phase to initiate parallel reduce tasks. These input and output files are stored in file system.

Input dataset is adapted by MapReduce framework from HDFS. Input dataset is taken as key-value pair and broken down for effective analysis. Then, the number of positive cases is mapped to corresponding diseases and shuffled accordingly. The number of positive cases is the reduced output. The final result is sent to the authorities and hospitals for preventive measures.

2.4. System architecture

In this work five MapReduce key functions are used to get the desired output/outcome. The functionalities of each functions are as follows:

Hadoop offers five daemons with each daemon possessing a Java Virtual Machine-

- Data Node
- Name Node
- Job Tracker
- Secondary Name Node
- Task- Tracker

Demons which store data and metadata, i.e.; DataNode and NameNode, come under a part of Hadoop Distributed File System (HDFS). The TaskTracker and JobTracker, which keep track and actually execute the job, come under MapReduce layer. The HDFS is used in this proposed research work because of the following reasons.

Large Dataset: Current population of India is more than 1.3 billion, if we want to analyze data for that amount of people then the dataset will be huge. That much huge amount of data can't be processed by normal file system. That's why to get a smooth workflow the HDFS is used for analysis.

Data Replication: For working with a large dataset, occurrences of unfortunate situations like hardware failure, crashing of a node are pretty common. In such situations data loss is occurred. To overcome this kind of problem HDFS is providing a feature called data replication. The data is copied across numerous nodes in the cluster by the creation of duplicates. This methodology is maintained across stipulated time intervals by HDFS and the duplication process is taken care of by the same. The moment as machine in the cluster crashes, the data should be retrieved from other machines. Loss of data is far sighted threat and almost negligible.

Scalability: Our main goal of the work is to analyze healthcare dataset using Hadoop and facilitate a smoother conduct of the fight against COVID-19. So, the proposed work is scalable in order make it a dynamic project. This is achieved by using HDFS. In HDFS the infrastructure is scaled up by adding more racks or clusters to this system.

Data Locality: In older systems, the data is brought at the application layer and then worked upon. In this proposed research work, as a consequence of the huge bulk of data, bringing data to the application layer has lowered down the overall performance.

In HDFS, the computation part is brought to the Data Nodes where data resides. Hence, with Hadoop HDFS, computation logic is not moved to the data, rather than data is moved to the computation logic.

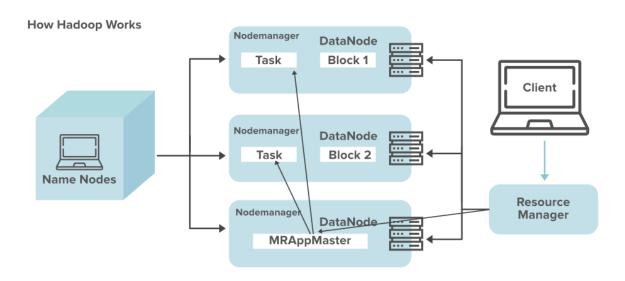


Figure 4: Daemons of Hadoop

CHAPTER 3 METHODOLOGY

3.1. Data Analysis (Any algorithm, queries or tools used)

Given below is a screenshot of our live retrieved dataset:

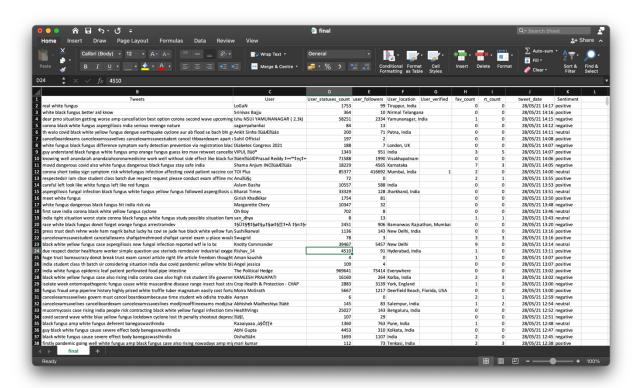


Figure 5: CSV format of our Dataset

3.1.1. Creating Twitter Application

1. Open the website dev.twitter.com/apps in the Browser.



Figure 6: Develop.Twitter

2. We will now see the website suggesting us to sign in. So, we sign into our twitter account.

Click on Create New App.



Figure 7: Creating new app

3. Now, scroll down and tick the option Yes, I agree and then click Create your Twitter application.

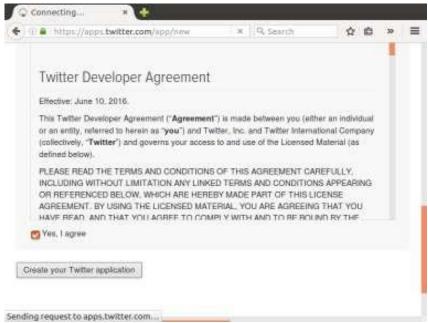


Figure 8: Agreement

4. Click on manage keys and access tokens.

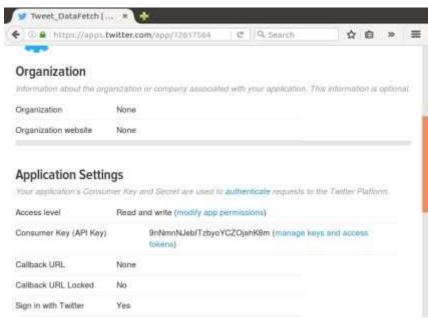


Figure 9: Manage Keys & Tokens

5. Now click on Create my access token.

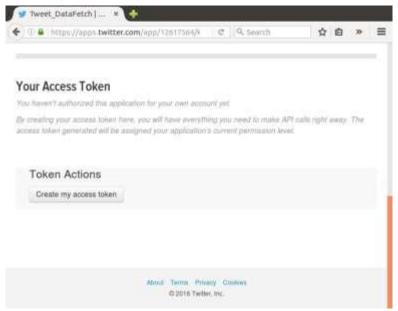


Figure 10: Create your own access tokens

3.1.2. Getting Data using Python

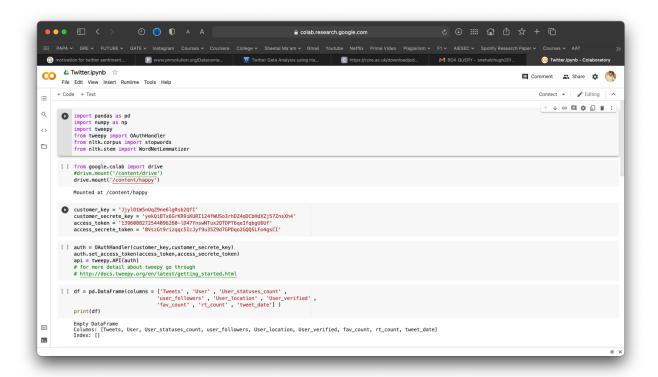


Figure 11: Getting Data using Python - 1

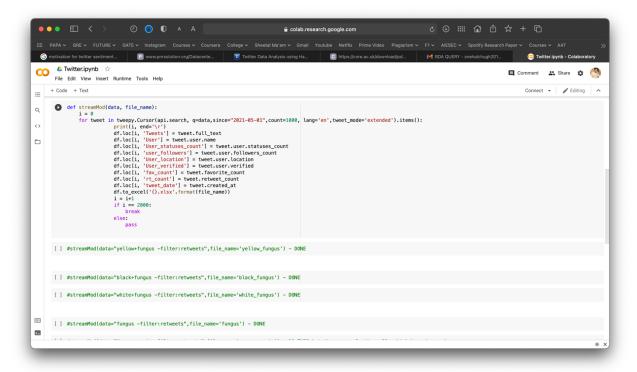


Figure 12: Getting Data using Python – 2

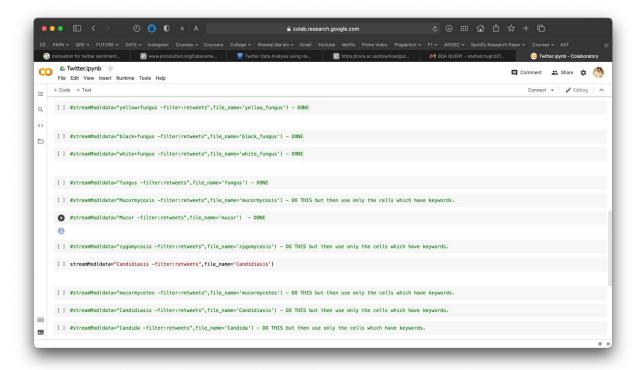


Figure 13: Getting Data using Python - 3

3.2. Hive queries

1. Create table tweets_exp.

```
CREATE EXTERNAL TABLE IF NOT EXISTS tweets_exp
(
id BIGINT,
tweet STRING,
user_name STRING,
user_statuses_count INT,
user_followers_count INT,
user_location STRING,
user_verified BOOLEAN,
tweet_fav_count INT,
retweet_count INT,
tweet_date STRING
```

2. Load the data in tweets table

load data inpath '/BDAFINAL/final_white_plus_yellow_final.csv' into TABLE tweets_exp;

```
A bigint bigint string string
```

Figure 14: Loading data into tweets table

3. Display the table

Select * from table tweets_exp limit 10;

OUTPUT:

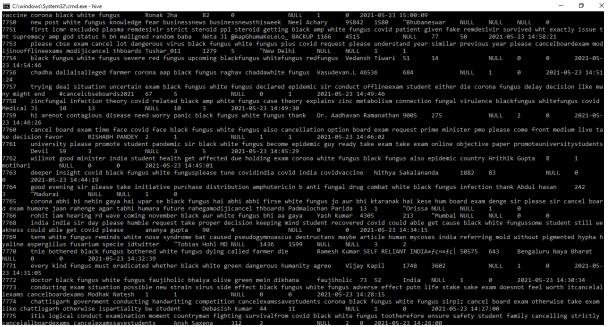


Figure 15: Displaying the table

4. Create table dictionary

```
CREATE EXTERNAL TABLE dictionary

(
type string,
length int,
word string,
pos string,
stemmed string,
polarity string
)

ROW FORMAT DELIMITED
FIELDS TERMINATED BY '\t';
```

5. DESC table

```
desc dictionary;
OK

type string
length int
word string
pos string
stemmed string
polarity string

Time taken: 0.315 seconds, Fetched: 6 row(s)
```

Figure 16: Description of table dictionary

6. Display the table dictionary

Select * from dictionary limit 10

```
Live select * from dictionary limit 10;
Live s
```

Figure 17: Viewing the dictionary table

7. Create view temp 1

create view temp_1 as select id, tweets_exp.tweet, words

from tweets_exp

lateral view explode(sentences(lower(tweet))) dummy as words;

```
Section of the property of the
```

Figure 18: View temp 1

8. Create view temp 2

create view temp_2 as select
id,
temp_1.tweet,
word
from temp_1
lateral view explode(words) dummy as word;

```
Complete the control of the control
```

Figure 19: View temp 2

9. Create view temp 3

```
create view temp_3 as select
id,
temp_2.tweet,
temp_2.word,
case s_d.polarity
when 'negative' then -1
when 'positive' then 1
else 0
end as polarity
from temp_2 left outer join dictionary s_d on temp_2.word = s_d.word;
```

```
New Stf Nive auto. convert. Join-false;
2021-06-14191-40212,55 INFO [main] org. apache. hadoop. hive.conf. HiveConf - Using the default value passed in for log id: 00e05d70-d561-db0c-b2db-07325f1c27c0 0 atol. 1419-40212,55 INFO [main] org. apache. hadoop. hive.conf. HiveConf - Updating thread name to 00e05d70-d561-db0c-b2db-07325f1c27c0 0 atol. 0201-06-14191-40212,750 INFO [08006370-d561-4b0c-b2db-07325f1c27c0 main] org. apache. hadoop. hive.conf. HiveConf - Using the default value passed in for log id: 00e05d70-d561-db0c-b2db-07325f1c27c0 main] org. apache. hadoop. hive.ql. session. SessionState - Resetting thread name to main not be seen that the second of the conformation of the second of t
```

Figure 20: View temp 3-1

Figure 21: View temp 3 -2

10. Create table sentiment

```
create table tweets_sentiment as select id, case when sum( polarity ) > 0 then 'positive' when sum( polarity ) < 0 then 'negative' else 'neutral' end as sentiment from temp_3 group by id;
```

```
The Claimboom Systemi2 condear- him 2021-06-14191-49:56, 2021-107 [09e06370-d561-4b0c-b2db-07325f1c27c0 main] org.apache.hadoop.hive.ql.session.SessionState - Resetting thread name to main live create table tweets, sentiment as select as the content of the cont
```

Figure 22: Creating sentiments table -1

```
Set Nive evec. reducers max-remainer)
In order to set a constant number of reducers:
set mapreduce.job.reducers.crumber?
2021-06-1419:43:15,310 WARN [09005070-d561-db0c-b2db-07325f1c27c0 main] org.apache.hadoop.metrics2.impl.MetricsSystemImpl - JobTracker metrics system already initialized in the control of the control
```

Figure 23: Creating sentiments table -2

```
AppReduce Dob Laumched:
Stage-Stage-1: HDFS Read: 13070245 HDFS Write: 0 SUCCESS
Stage-Stage-2: HDFS Read: 13019444 HDFS Write: 149508 SUCCESS
Total RapReduce CFU Time Spent: 0 msc
Total RapReduce CFU Time Spent: 1 msc
Total RapReduce CFU Time Spent: 0 msc
Tot
```

Figure 24: Creating sentiments table - 3

11. Output with tweets and id

```
create table tweet_sentiment as select id,tweet, case when sum( polarity ) > 0 then 'positive' when sum( polarity ) < 0 then 'negative' else 'neutral' end as sentiment from temp_3 group by id,tweet;
```

```
201-80-1419-4619,800 NPO [09e85d70-d561-4b0c-b2db-07325f1c27c0 main] org.apache.hadoop.hive.conf.HiveConf - Using the default value passed in for log id: 09e85d70-d561-4b0c-b2db-07325f1c27c0 main] org.apache.hadoop.hive.ql.session.SessionState - Resetting thread name to main live constructions of the construction of the cons
```

Figure 25: Output with tweets & ID - 1

Figure 26: Output with tweets & ID - 2

3.3. Visualization screenshots

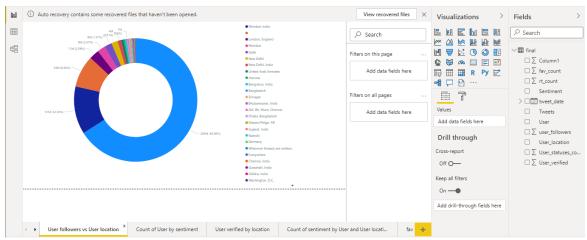


Figure 27: User Followers vs User Location

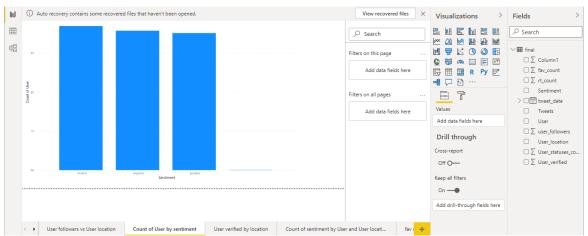


Figure 28: Count of User by Sentiment

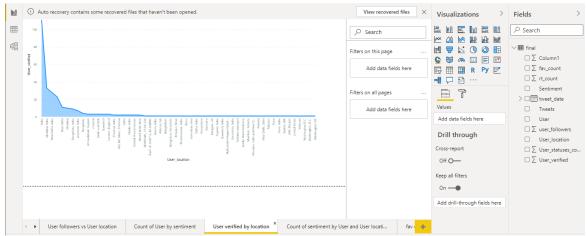


Figure 29: User verified by Location

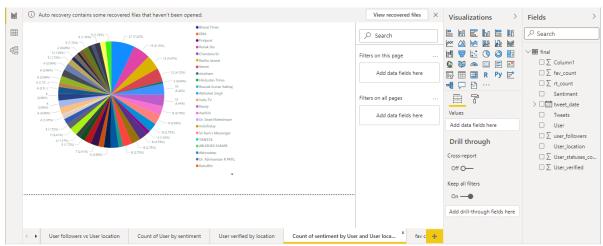


Figure 30: Count of Sentiment by User and User Location

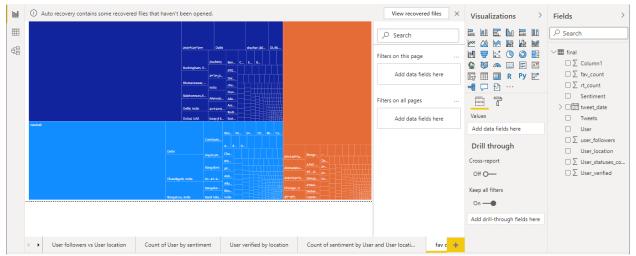


Figure 31: Favorite Count by sentiment and user location



Figure 32: Retweet Count by Sentiment

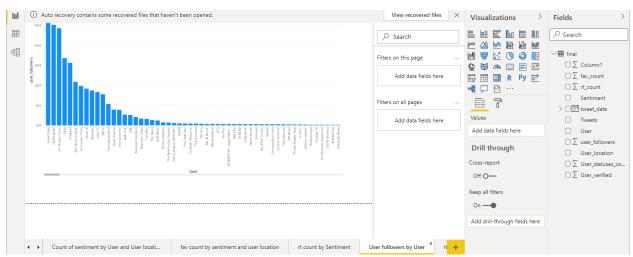


Figure 33: User Followers by User

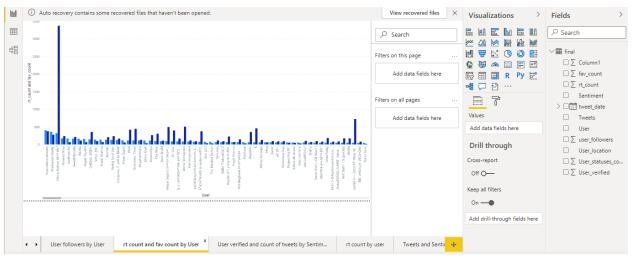


Figure 34: Retweet count and Favorite count by User

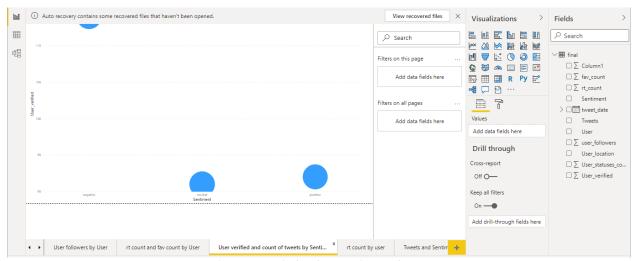


Figure 35: User verified and count of tweets by Sentiment

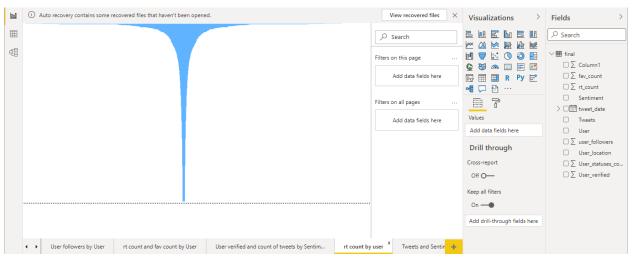


Figure 36: Retweet Count by User

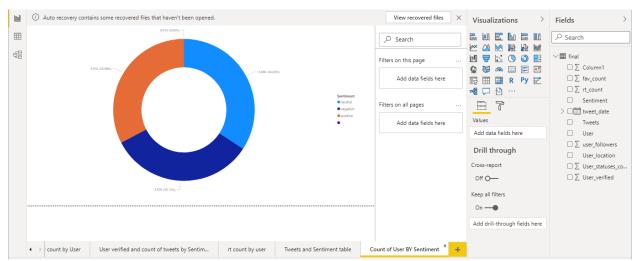


Figure 37: Count of User by Sentiment

CHAPTER 4 FUTURE SCOPE

4.1. Conclusion

This study focuses on the application of big data analytics technology in the prevention and control of major public health incidents. First of all, we clarified the definition, characteristics, and prevention difficulties of major public health incidents. To cope with these difficulties, the use of big data is an important means to assist prevention and control in major public health incidents. Governments may make full use of the application of big data in an epidemic situation in all aspects of prevention and control, and they can further improve the epidemic prevention mechanism based on big data analytics. In terms of information collection, data collection platforms for the Internet of Things, mobile devices, navigation and search engines, social media, and large-scale gene banks can be fully established. On the basis of information collection, it is necessary to establish an early warning detection mechanism for big data analytics, e.g., using visual analysis, deep learning, and forecast line analysis techniques. This may be used as a basis for early warning and forecasting, formulating plans, rapid decision-making, and starting emergency mechanisms.

Second, governments can further improve their epidemic response mechanisms based on big data analytics. Big data technology can be used for diseases identification, decision support, coordination and communication, and technical support. Diseases identification typically uses predictive analysis of infectious disease dynamic models combined with data to make predictions regarding the criticality of an event, provide support for management decisions, report, and take timely response measures. Graph database analysis and geographic information systems can provide a significant advantage in tracking infected persons and their contacts, thus determining the source of infection. For research into virus sources and the research and development of specific drugs and vaccines, we should fully utilize the potential of big data technology for research support and technical consultation regarding genetic data and real-time patient data transmitted by the Internet of Things.

Third, the government should establish an epidemic repair mechanism based on big data analysis and promoting the sharing of big data in different regions, industries, and platforms. This includes the use of big data to eliminate fear, for recovery, audit assessment, and policy adjustment. Big data analytics can be convenient in ameliorating public fear by revealing the real-time epidemic situation and clarifying rumours. The big data analysis model can also be used to estimate the impact of the epidemic on political, economic, and social development, so as to assist governments to make suitable decisions, make policy adjustments, integrate prevention and resistance measures, and promote a rapid economic recovery.

It is important that big data analytics is merely a supportive method to assist in ex-ante prediction and ex-post prevention and control. Big data analytics has certain limitations and application premises. For instance, in terms of predicting in advance, some methods of early warning with big data may be directly used for diseases that scientists already understand, such as influenza, because relevant massive data have been accumulated. However, faced with the first generation of new viruses & diseases, it may not be possible to directly generate closely related big data, so it may not be possible to use big data technology for immediate analysis. Notwithstanding, researchers may explore the combination of data from other relevant sources for analysis, such as the incidence of internal and external causes (climates, other epidemics, etc.) that could affect how a virus is produced, and the probability of the occurrence of new viruses being observed and analysed in advance. Moreover, the application of big data in epidemic prevention and control must consider the practicability of administrative rights, privacy protection, cost, and so on, and the balance of interests with public epidemic prevention so as to ensure operability.

4.2. Future Work

Future work involves collaborating with "Spark NLP," an open-source library for the processing of natural languages, developed on top of Apache Spark and Spark ML. Integrating with ML Pipelines offers a simple API. The Spark NLP library, which contains Scala and Python APIs for use by Spark, is written in Scala. It has no reliance on any other library of NLP or ML. The library includes the ability to train, modify and store models as a native extension of the Spark ML API so that they can run on a cluster or other machines or save for later.

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