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Neeta Niraula Saint Cloud State University, nine1401@stcloudstate.edu

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Web Log Data Analysis: Converting Unstructured Web Log Data into Structured Data Using Apache Pig

by

Neeta Niraula

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Abstract

Data extraction and analysis have recently received significant attention due to the evolution of social media and large volume of data available in an unstructured form. Hadoop and MapReduce have been continuously implementing and analyzing large amount of data. In this paper Apache Pig, which is one of the high-level platform for analyzing large volume of data and runs on the top of Hadoop is used to analyze unstructured log files and extract information. In this paper, weblog server files are used to analyze and extract meaningful information in an unstructured form to a structured form in Apache Pig framework

The main purpose of this paper is to extract, transform and load unstructured data in an Apache Pig framework and analyze the data and its performance on local mode as well as MapReduce mode. This paper further explains in brief about the different steps required to analyze unstructured web server log files in Apache Pig. This paper also compares the efficiency when a large volume of data is processed on MapReduce mode and local mode.

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Chapter 1: Introduction

In fact, data is always crucial for business professionals within an enterprise to study business details, and the rise of big data has now further facilitated many other areas to explore among big business enterprises. Among these areas, one of the major area involves analysis and organization of different types of data collected in the business enterprises. Data is mainly distinguished into two different types: unstructured and structured data in an organization. Despite the fact, these data types are available in different formats and are likely to be managed differently, the role of each data type is potentially significant for an organization to keep records and make impactful business decisions.

Data is growing at the rate of 50% per year and is usually represented in fivedimensional models: volume, velocity, variety, veracity, and value, represented as 5V. Data analyst and business stakeholders rely on these data types to produce some significant results on collecting, archiving and discovering data to some useful end results.

Structured Data

Structured data refers to those data types which is available in a highly organized form. They are sometimes called traditional data and is simple to enter, store, query, and analyze. Unstructured data can be easily represented in the form of particular tables or schemas. Before the era of big data, structured data was what enterprises used for making business decisions. Structured data is defined in terms of schema and tables and is used to analyze using structured query language or excel spreadsheets to perform queries with relational database. Structured data analysis tools and techniques is developed with the improvement of data processing by computers, lowered storage cost and new format of data. However, many data-driven companies throughout the world are nowadays beginning to take emerging data sources seriously and both structured and unstructured data are consulted, queried, assimilated and improved to make potential business decisions.

Unstructured Data

Nowadays, unstructured data is growing continually due to the increase in data storage platforms and the infinite number of complex data sources such as social media platforms, mobile application, health data, weblogs, emails, etc. Most of the business interactions now, in fact, is unstructured in nature. The fundamental challenge on managing unstructured data is the diversity in data. Unlike structured data, unstructured data need to work with some specialized tool and be structured in order to analyze.

Qazi and Sher (2016) states, "The world creates 2.5 Quintillion Bytes of data per day from unstructured data sources like sensors, social media posts, and digital photos". The unstructured data doubles every three months and if left unmanaged the sheer volume of unstructured data will be big (*Insights on governance, risk and compliance*, 2010). In the world of Big data, the accelerating growth of data sources, volumes and configurations are mainly driven due to unstructured data. Therefore, organizations are now little more aware of the volume, composition, risk and business value of their unstructured data. Data warehouse enterprises, like IBM, now have started using big data tools and techniques like, Hadoop, Map-Reduce, etc. for analyzing all data types, which were previously only used in very specific areas to analyze the opportunities inside unstructured data. Therefore, given these challenges for managing unstructured data, it seems reasonable to accept the challenge and investigate and analyze unstructured data to extract some useful information from it using one of the big data analytics tool, Apache Pig, which is currently trending throughout the global market.

Apache Pig

Apache Pig, an open source high-level data flow system, is an abstraction layer on top of Map-Reduce and performs effectively on any data size, type or location. This methodology is popular and implemented by Yahoo since 2006 to have an ad-hoc way of creating and executing Map-Reduce jobs on very large data sets. It appears that Pig challenges to be better than RDBMS, DBMS based on performance, storage, and transaction level fault tolerance (Minsker, 2015). Although there are other different big data analytical techniques that are used for data analysis, Apache Pig has proved itself to be most efficient for analyzing unstructured data.

Pig uses ETL process for data warehousing. ETL, which stands for "extract, transform and load" is the set of functions combined in one tool and is used to extract large amount of data from numerous database. ETL process involves loading real-time data without interrupting data processing or conducting analysis that supports decision-making process. A common use case of ETL includes converting CSV files to format readable by relational database where user's input CSV-like data files and import it into a database.

Another advantage of Pig is that it can easily work on raw data. Unlike other big data analytics tools, Pig is the most efficient tool that can work on any sort of unstructured, semistructured, and structured data. Pig can easily load and process data using some user-defined functions that can be translated into a series of Map-Reduce jobs to run on an Apache Hadoop Cluster.

Although, the concept of data mining and processing has been outraged for years, extracting and analyzing raw data is always a question of challenge among data analytics. The performance, time, complexity, and the accuracy of data is vital and is needed for every business enterprise to move forward. Therefore, further research is needed that will elaborate other different findings to analyze unstructured data and extract some useful information to accelerate business plan strategies from the extracted data.

Research Questions

In this paper, the data in various unstructured forms are analyzed, examined, and converted into structured form using Apache Pig. Specifically, the following questions will be researched:

- 1. How to extract and transform unstructured data into a structured form using Apache Pig analysis tool?
- 2. How does Apache Pig, works on the top of MapReduce?
- 3. How to manipulate unstructured data to extract various information using Pig processing system?
- 4. How meaningful information can be generated from the unstructured data and how does it helps to make useful business decisions?

In the course of answering these questions, some unstructured weblog files is being used as an input and generated in a structured format using Pig processing system. The unstructured log files are extracted and stored in Hadoop Distributed File System, HDFS.

Scope of Study

To keep study compatible and functional, following parameters are used in this research:

Server Operating System: The server operating system selected for this project is Ubuntu 16.0.1. It is selected because several features of Apache Pig and Hadoop are upgraded in this version of Ubuntu.

Processing System: The processing system selected for this research is, Apache Pig. It is selected because Apache Pig supports unstructured data more efficiently than any other big data analytics tools.

Database Management System: The database management system selected for this research is Hadoop Distributed File System, HDFS. It is designed to store very large data sets reliably. The main reason for using HDFS in this research is that HDFS allows us to store filesystem metadata and application data separately within the same system.

Type and Size of Files: Unstructured data is used as input. More specifically, this research will be focused on using web server log files varying from few megabytes to gigabytes and generating results in a structured format as output.

These limitations made the study manageable in scope and will hopefully make it easier for the reader to evaluate and use the results.

Chapter 2: Literature Review

The review of literature can be broken into four parts. First, the advantage of Hadoop data analytics tool of extracting information from various data types is discussed. Second, importance of unstructured data for making business decisions is described. Third, various Hadoop analytics tool is discussed based on their functionality. Last, how Hadoop analytics tools has been successful in getting all the required information throughout the distributed data is discussed.

Minsker (2015) stated that the database's greatest power lies in the ability to process and store data that was not possible to be analyzed together due to its volume and unstructured form. She further stated that Hadoop has opened the door to analysis sentiments of society on social media providing them with the path of understanding the positively and negatively charged communications that take place throughout the social media. Lai, Chen, Wu, and Obaidat (2013) stated that Hadoop can gain response at a higher speed on implementing parallel processing over many nodes. When one node fails to work, Hadoop can obtain backup in no time and can recover data for any type of format, also known as fault tolerance. Francis and Kurian (2015) stated that Hadoop implements Map-Reduce framework which helps to retrieve and process unstructured data and map into rows and columns to a form a structured database. Map-Reduce framework can pair data as a key value pair and can be easily implemented using Apache Pig or Hive. Hive operates on the server side whereas Pig operates on the client side for data processing.

According to Quer and Sher (2016), big data is an opportunity that not only allows to deliver competitive advantages per data solutions but has also built an ecosystem throughout

the globe to improve technological power. According to the article ("Insights on governance, risk, and compliance", 2010), the conventional relational database cannot handle unstructured data and hence framework like Hadoop has been introduced to process unstructured as well as structured data in a high-speed platform and perform a more comprehensive analysis on big data using distributed and parallel processing systems. Holzinger and Pasi (2013) stated that about 80% of data within organizations are unstructured and unfit for traditional processing. Therefore, using big data will enable the processing of unstructured data and increase system intelligence that can result in different opportunities to study business data such as, improving sales, increasing understanding on customer needs, supporting marketing initiatives and fraud monitoring. "Organisations all need to realize that every data has value", states Al Almond, head of U.S. privacy and social media compliance at TD Bank.

Elizabeth (2013) stated that the analysis of unstructured data in business helps to discover current topics about the products from customer opinion. She also stated that the generated data in any form can be useful to find patterns in reports that may seem to predict business related activities. Business professionals can further discover and understand previously unknown issues and concerns from public feedback on different social media, social forums, etc. Herzig (2011) stated that both unstructured and structured data types are complimentary for business data and hence hybrid queries need to be implemented in order to improve data analysis. Beach and Schiefelbein (2013) stated that monitoring log files, text messages, and online customer product reviews can be the most effective medium for an organization to identify hidden risk before they emerge as full-blown crisis. They further justified that checking all types of data can provide deeper understanding of ongoing business activities, ensure compliance with certain regulations and monitor employees' engagement and retention. According to hadoop.apache.org, big companies like IBM, Amazon, Google, Facebook, AT&T, NBCUniversal, FedEx, etc. has already adopted Hadoop to analyze data for financial, marketing, advertising, and sentiment and risk analysis purpose.

Gupta and Kiran (2014) stated that the tool used by Big Data Analytics for processing unstructured data is, Hadoop. Hadoop Implements Map-Reduce framework and can be used to filter unstructured data and semi-structured data into structured formats that can be loaded into any other analytic platforms. Hadoop consists of two components, Hadoop Map-Reduce for parallel data processing to extract higher-value data from raw files and the Hadoop distributed file system (HDFS) that supports low-cost, scale-out storage. Hadoop is much more effective than any other data analytics tools and it provides flexibility that can store any data type of any size. Hadoop provides scalability that can store data from terabytes to petabytes. Hadoop is designed to work on high volume, high velocity, and on various varieties of data. Therefore, Hadoop ecosystem is strong enough to combine any type of old and new data sets in different powerful ways. Devakunchari (2014) stated that Hadoop is designed to run on a large number of commodity servers. The servers are arranged in a system and Hadoop software runs on every server. He further stated that Hadoop is best suitable to run large datasets that are complex and computationally extensive. Nandimath et al. (2013) stated that Hadoop ecosystem consists of four main components, Hadoop Common, Hadoop distributed file system, Hadoop Map-Reduce, and Yarn. Hadoop Common also known as Hadoop core contains utilities and libraries that can be used by other modules within the

Hadoop ecosystem. It contains necessary java archives files and scripts required to start Hadoop.

Hadoop distributed file system is the default storage for data processed inside Hadoop. It creates several replicas of the data block across different Hadoop clusters to make data accessible and reliable. HDFS architecture works on master-slave model and consists of three main components, NameNode, DataNode, and Secondary NameNode. NameNode is the master node to keep track of the storage clusters and DataNode acts as a slave node performing the data processing within the Hadoop cluster. They further explained that Map-Reduce is a java-based system where actual data from HDFS gets processed. Map-Reduce further breaks down the big chunk of data into smaller sub-tasks where Map sends query input into various clusters for processing and Reduce collects all the processed data as a single output. Meanwhile, both input and output task are stored in HDFS. Another main component in Hadoop ecosystem is Yarn which is responsible for dynamic resource utilization in Hadoop framework. According to Horton works, Yarn has extended the Hadoop capabilities to adopt compelling new technologies within data center and is very cost effecting. It further provides a consistent framework to write data access applications that run in Hadoop. These are the core components in any basic Hadoop framework but there are several other components that form an integral part of the Hadoop ecosystem with the intent of enhancing the power of Apache Hadoop and providing better integration with databases.

Hadoop provides simplified access to the data stored in HDFS and sends data to data processing layer using different types of infrastructures like Hive, Pig, Mahout, Avro, etc. All of these infrastructures resides on the top of the Hadoop ecosystem to summarize the big data concepts. Hive was originally developed by Facebook and later by apache that allows users to write query in a SQL-like language, HiveQL, converting to Map-Reduce. It is so simple that any developer having the knowledge in SQL can write query for data processing. It first stores schema metadata in database and stores the data in HDFS and operates on the server side of cluster ("Hadoop ecosystem: An introduction", 2016). Gates et al. (2009) states that Pig acts as a high-level data flow in between SQL and Map-Reduce. Pig was first adopted by Yahoo and has its own programming language known as, Pig Latin which can be compiled into a sequence of Map-Reduce jobs and executed in Hadoop ecosystem. Pig is a procedural programming language and supports all parallel programming conditional constructs (FOREACH, FLATTEN, GROUPBY, etc.). Gates et al. (2009) further admitted that Pig is a very useful framework for processing log files, aggregation of data warehousing, filtering media files, etc. Eluri, Ramesh, Al-Jabri, and Jane (2016) stated that in order to elevate the scalability of data clusters, Apache Mahout data clustering algorithms can be implemented on the top of Hadoop using Map-Reduce paradigm. They studied two different clustering techniques K-means clustering technique and Canopy-clustering technique using Apache Mahout. The K-means thus seems to be efficient to implement but is only suitable for globular data set whereas the Canopy clustering technique implementing Mahout was suitable for both globular and non-globular data set. Apache Mahout is, therefore, useful on filling the data from large data clusters by using recommendation engine of Apache mahout.

Jain (2013) stated that using Apache Sqoop data can be efficiently transferred between Apache Hadoop and structured data store relational databases. Sqoop is useful when there is a need of loading bulk data into Hadoop from production systems or accessing it from MapReduce applications running on a large cluster. Sqoop is a batch-oriented and thus is not suitable for low latency interactive query operations but mitigates excessive data load when required.

Information related to this research has been reviewed and studied in this chapter. In the next chapter we will focus on the research methodology applied in this research and the achieved results.

Chapter 3: Methodology and Results

As stated in earlier chapter about Hadoop ecosystem and its component, this research adopts one of the Hadoop components, Apache Pig as the research topic and unstructured data, especially some log files over the Hadoop framework. The log files is processed on Apache Pig framework and analyzed using Apache Pig scripting language, Pig Latin to extract useful information in a structured form. Output is based on the type of input log files used and the Apache Pig framework where the data is being processed and analyzed. Map-Reduce allows programmer to specify map function on the input data followed by reduce function to generate output. However, working on how we to fit any data to work under this pattern is itself a big challenge. In Pig, the data-structure itself are multivalued and nested thus can process any data type. Pig basically consists of two different components, Pig Latin and Grunt. Pig Latin is the scripting language that Pig uses for series of extractions, and transformations to produce output from the input on Pig environment known as Grunt. The development cycle in Map-Reduce framework is tedious and hectic due to the codes being too long and complex. In Pig Latin just using few lines of code in a simpler way can do the same job. Pig was created by Yahoo to provide data analytics to mine large amount of data. Pig divides the data transformation into series of Map-Reduce job within few lines of code which allows developers to focus on data rather than nature of execution. Pig automatically optimizes task without execution Pig Latin, on writing query supports most of the commands similar to SQL and relational operators. Pig scans large volume of dataset once which might not be suitable when we need to process data on batches. But Pig runs on client side

applications and launches jobs that interacts with HDFS from any workstation. Pig can also be implemented for data profiling using sampling and also for quick hypothesis testing.

Pig Architecture

Pig performs on the top of Hadoop and can read data from HDFS for data extract, transform and load (ETL) process. In this section, we will discuss in detail about the Pig architecture and its components. Pig uses Pig Latin as its scripting language that be further written using built-in operators that runs on Pig environment. There are mainly three ways to execute Pig Latin scripts: first, Grunt mode which is an interactive mode of Pig. In addition to this, there are certain useful shell and utility commands supported by grunt shell useful for testing syntax and ad-hoc data exploration. Second, Script mode which runs as a set of instructions from a file and is executed by the Pig server. Third is the embedded mode certain user defined functions (UDF) can be used using different other languages like Java, Ruby, Python, etc. This mode is suitable to create Pig Scripts on the fly.

As shown in Figure 1, Pig scripts from Grunt or Pig server passes through Parser. The parser parses the code by checking syntax in the script and generates a direct acrylic graph (DAG) as an output. DAG represents all the Pig Latin statements and logical operators. The logical operator acts as nodes and data flows in between the edges.

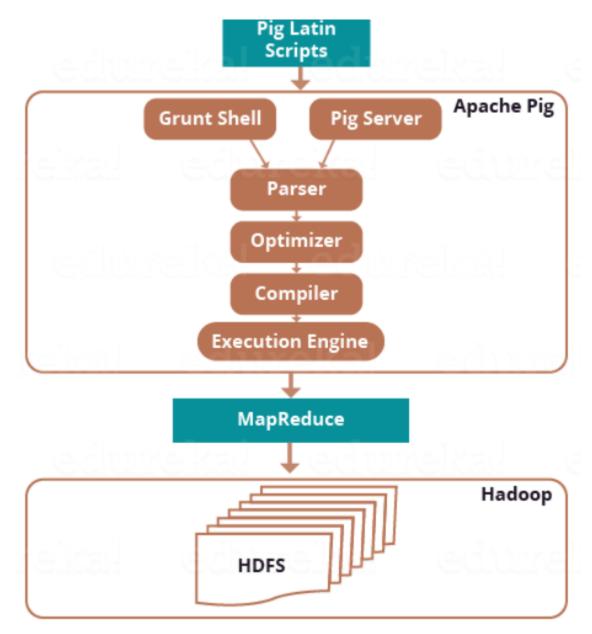


Figure 1. Pig architecture ("Hadoop ecosystem: An introduction", 2016).

The Optimizer gets DAG as the input and optimizes the script using splits, merge, transform, reorder, etc. The optimizer also reduces the amount of data in the pipeline or reduce phase whereas for most of the Map-Reduce layer optimization rule aims to optimize

the MapReduce job properties either to include compiler of not. Additionally, at this layer query Joins, ORDER BY and GROUP BY functions are implemented.

After the script optimization, the compiler compiles the optimized code into series of MapReduce Jobs. Pig jobs can be automatically converted into MapReduce jobs by compiler. When the MapReduce jobs are submitted for execution to the execution engine either in local mode using single JVM, required result are generated as output. To view the result *DUMP* statement is required.

Pig Data Types

There are two different types of Pig's data, scalar types (single value) and complex types. Scalar data types are represented in Pig interfaces by java.lang classes and are same as the types in most programming languages. Complex data types are Tuples, Bag, and Map. A separate table is provided to explain data model in Pig as shown in Table 1.

Table 1

Example of Pig Data Model

| Id | First Name | Last Name | Faculty |] |
|--------|------------|-----------|------------|-----|
| 1 | John | Doe | Management | ු ව |
| 2 | Mary | Smith | IT | Bag |
| 3 | Jane | Porter | Education | |
| Tuples | | | | -] |

A Tuple is a fixed length, ordered collection of fields and also represented as the record stored in relational database. For each table, we can access fields in each tuple using indexes of the fields. The fields elements can be of any type and is analogous to the the row in SQL. Fields are compared as SQL columns and referred by positions. This allows Pig to check the data in the tuple. Pig further allows user to reference the fields of the tuple by name.

Tuple constants use parentheses to indicate the tuple and commas to delimit fields in the tuple as shown in Example 1.

Example 1. {1, John, Doe, Management}

Similarly, a Bag is an unordered collection of a set of tuples. A bag might be redundant which means, it can have same tuples within the schema and hence not reference by the position. For Apache Pig to process a bag, the sequence for fields and the respective data types must be in sequence. Bag constants are constructed using braces, with tuples in the bag separated by commas as shown in Example 2. which shows the relation between student and their faculty and constructs a bag combined from many ordered tuples. In Example 2, John, Doe, Management), (Mary, Smith, IT), (Jane, Porter, Education) are tuples and Bag is represented by parenthesis including all these tuples.

Example 2. {(John, Doe, Management), (Mary, Smith, IT), (Jane, Porter, Education)},

{(John, Doe, Management), (Mary, Smith), (IT, Jane, Porter), (Porter, Education)}

(John, {(Doe, Management), (Mary, Management)}).

A map is a key-value pair represented as data elements. Maps contain unique keys and are represented as chararray [] and contain unique column name that can be indexed to access the value associated with it. Because Pig does not know the type of the value, it will assume that it is a byte array. However, the actual value might be something different. If the value is of type other than the byte array, Pig will figure out the value of the data type at runtime . As shown in Example 3, Map constants are formed using brackets that delimits the map using *hash* between keys and values, and a comma in between key-value pairs is used. In

Example 3, there are two keys, "FirstName" and "Id". The first value is a char array and the second is an integer.

Example 3. [FirstName# John, Id#1], [FirstName#Mary, Id#2] Both simple and complex data types can be attached to the Pig schema during Load. By default, char array is the default data type for Pig.

Besides of the different Pig invocations, Pig mainly runs on two modes, Local and MapReduce mode. In this research, we will be running Pig on MapReduce mode where Pig job runs as a series of map-reduce jobs and HDFS is the storage for both input and output data file. Different types of log files are processed in Pig processing system such as web server logs files and access log files.

General Workflow of Apache Pig

When user-defined functions is implemented or any input log files is used inside the Pig GRUNT shell, the following steps are involved as explained in Figure 2.

- 1. Script Parsing
 - Check the syntax and valid reference variables.
 - Check whether data types are properly initialized or not.
 - Schema inference
- 2. Logical Optimizer
 - Pass logical plan by generating a sequence of well-founded semantics query models to solve the query.

- 3. Physical plan
 - Translate logical plan to physical plan for each logical operator specifically by describing the physical operators that Pig will use to execute the scripts.
- 4. MapReduce plan and its optimization
 - Assign each physical operator to a MapReduce stage (Map task and a Reduce task).
 - Minimize the number of reduce stages based on the nature of operations of MapReduce optimizations.

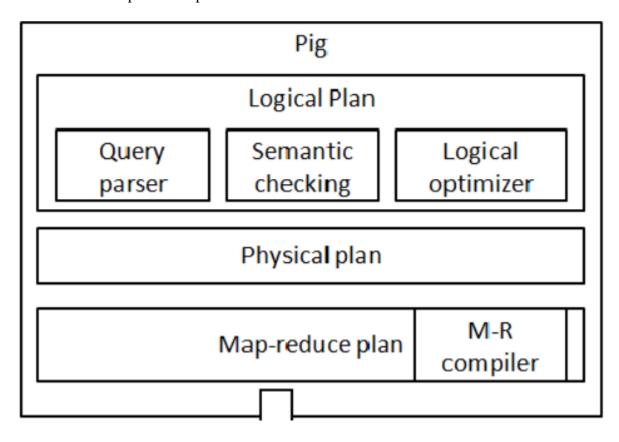


Figure 2. Workflow architecture of Apache Pig ("Hadoop ecosystem: An introduction", 2016).

The processes involved inside the Pig architecture to transform the Pig Latin programs into executable code is explained in above Figure 2. The executable code is in the form of MapReduce tasks passing through different stages that includes a parser, logical optimizer, Physical optimizer, MapReduce optimizer and the MapReduce compiler.

Pig can also be used for column transformation, filtering, ordering, and Custom aggregation. Pig also opens up the power of MapReduce. To know the workflow or Apache Pig, our research is focused on analyzing data from web log files as explained in the test case below.

Case Study: Access Log Report Analytics

This case study includes using apache Pig commands and creating some user defined functions to process access log data files. The log reports used in analysis are the two month's HTTP requests to the NASA Kennedy Space Center WWW server in Florida . The log reports contains logs collected from July 1, 1995, through July 31, 1995, a total of 31 days. The log file contains Server IP address, Method (GET / POST), Request URI (request_link and request destination) , HTTP status code, and User-agent. The case study illustrates the methodology of processing and analytics of Apache Pig using Hadoop distributed file system.

Problem. The format of access logs are very cryptic. These files exist typically for technical site auditing and troubleshooting. In order to maintain security and investigate the incidents around the web within any organizations, these logs need to be monitored and analyzed. With proper analysis and inspection, one can identify intrusion attempts, misconfigured equipments, user behavior, and much more.

In this case, we will find the total number of times a website that has been visited in

last 31 days (July 1 to July 31).

Input (Access log files). Access log files are the files generated by server and also

known as server log files to provide information requested to the server from users. When an

user connects to a site, the computer, browser, and network will deliver some data to the site

server itself to create a record that file was requested. A sample of web server log file appears

below as Figure 3.

| | d X |
|--|---|
| File Edit Format View Help | |
| <pre>19 7.281.55 - [01/Jul/1995:00:00:14 -0400] "GET /hittpr/Applio/ HTTP/1.0" 200 4250nicomp6.unicomp.net - [01/Jul/1995:00:00:26 -0400] "GET /inuges/launchpalmas-mall.gif HTTP/1.0" 200 4173205.189.154.54 - [01/Jul/1995:00:00:24 -0400 "GET /inuges/MASA-logosmall.gif HTTP/1.0" 200 7861x-orl2-01.ix.netCom.com - [01/Jul/1995:00:00:24 -0400] "GET /inuges/launchpalmas-mall.gif HTTP/1.0" 200 1173205.189.154.54 - [01/Jul/1995:00:00:24 -0400] "GET /inuges/MASA-logosmall.gif HTTP/1.0" 200 7861x-orl2-01.ix.netCom.com - [01/Jul/1995:00:00:24 -0400] "GET /shuttle/mission/sts-71/inuges/MaSA-logosmall.gif HTTP/1.0" 200 7177.200 915209 55.189.154.54 - [01/Jul/1995:00:01:08 -0400] "GET /shuttle/mission/sts-71/inuges/inages.html HTTP/1.0" 200 7634ww.al.proxy.aol.com - [01/Jul/1995 55.189.154.54 - [01/Jul/1995:00:01:08 -0400] "GET /shuttle/mission/sts-71/inuges/MaSA-logosmall.gif HTTP/1.0" 200 774/inu697.13/inages/KCS- mages/launchmedium.gif HTTP/1.0" 200 11853311p1.yab.com - [01/Jul/1995:00:01:29 -0400] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 200 774/inu697.5.0%112 - [01/Jul/1995:00:01:57 -0400] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 200 774/inu697.5.0%111995:00:01:57 -0400] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 200 774/inu697.5.0%12 - [01/Jul/1995:00:01:57 -0400] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 304 0netport-27.iu.net - [01/Jul/195:00:02:10 -0400] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 200 5071x-01400 "GET /shuttle/mission/sts-74/040] "GET /shuttle/mission/sts-74/inages/MaSA-logosmall.gif HTTP/1.0" 200 627611mmith.tercat.com - [01/Jul/1955:00:02:16 -0400] "GET /shuttle/mission/sts-74/54/inages/MaSA-logosmall.gif HTTP/1.0" 200 62740-shut-shut-shut-shut-shuttle/mission/sts-74/inages/KSC-logosmall.gif HTTP/1.0" 200 2740/inages/MaSA-logosmall.gif HTTP/1.0" 200 2740/inages/MaSA-0400] "GET /shuttle/mission/sts-74/inages/MaSA-0400] "GET /shuttle/mission/sts-74/inages/MaSA-0400] "GET /shuttle/mission</pre> | <pre>r /imagg ''GET', 200 398% 200 398% 200 398% 2055C-04% 169911: 1.0° 20% 5/WORLD-2. 2.gif H'' 295:00:(3:00.5/31 0.3985d0; 3:02:39 :00:02:4 .c.south: 00:03:0% 2:00'' 00'' 00'' 00'' 00'''' 00''' 00''' 00''' 00''' 00''' 00''' 00''' 00''' 00''' 00'''' 00'''' 00'''' 00'''' 00'''' 00'''' 00''''''</pre> |

Figure 3. Web server log files.

The data as shown in above Figure 3 is space separated. It consist of IP address,

timestamp, timezone, request type, requested link, request details, response code and user

agent(bytes). Usually, the scale of these files is quite huge and running queries in

conventional method is not possible. Therefore we will implement Apache Pig to manipulate

these log files and generate necessary statistics which helps us to understand the usage of the

server, website popularity, user visit frequency, and total bytes transferred. The main purpose of using access log files is to:

- Identify the most popular websites to get insight using user defined functions in Pig Processing system.
- ii. Implement custom load functions to load Apache's common log format files into Apache Pig.
- iii. Use several join operators from Pig join based datasets on values common to each dataset

Expected output. The result should be able to show the total number of visits of any website within the given timestamp.

Example 4. <u>www.google.com</u> has been visited 272 times from 03/15/2017 to 03/20/2017.

Solution. The logs files are downloaded from NASA-HTTP website. In order to process unstructured web server log files, the primary step is to load the datasets into Hadoop Distributed File System (HDFS). It is done by creating a directory in HDFS and storing the access log files in HDFS for further processing.

The primary objective of this research is to implement Pig processing system to find a structured records of the URL visits throughout any given time period recorded in an unstructured data type log file. The input data from HDFS is loaded into Pig system. Pig has a special keyword called *"Load"* to load data into PigStorage. In the Pig system, data is stored, and transformed by using relations. Each transformation of data is assigned to a variable as a data as shown in the example below. The location to load file is based on that particular mode

(MapReduce or local) in which Pig is executed. We implemented both *MapReduce* mode and *Local* mode. The MapReduce mode is also known as cluster mode where the file is located inside the specified HDFS location. In given Code Snippet 1, the data is loaded from the Hadoop Distributed File System located in '/user/AccessLogAnalysis/WebLogs/ AccessLogFile1' file location by using the variable '*data*'. The data is loaded specifying specific schema for the *AccessLogFile1* file. Let's take a look at ip_add[chararray], ip_add is the input keyword informing Pig that is of a chararray type.

The relations are defined for each transformed data which act as a reference to a set of data. Ip_add is reference as the IP address of all the URL listed in the input *AccessLogFile1* file. In other words, these relations acts as a set of tables with user-defined functions and columns. As mentioned in Code Snippet 1, the data loaded as input is space separated and each line is a distinct event. Output generated by *Describe* command confirms same schema which is specified along *with LOAD*. After the data is loaded as shown in Code Snippet 1, in order to calculate the total bytes transferred in every interval of time, the data is aggregated based on upon the timestamp as shown in Code Snippet 2.

```
Code Snippet 2.
time_data = GROUP data BY timestamp;
DESCRIBE time_data;
byte_count = FOREACH time_data GENERATE group AS
timestamp, SUM(data.bytes) AS total bytes;
```

The above Code Snippet 2, defines a relation 'time_data' and 'byte_count'. The

time_data consists of data grouped by *timestamp*. So, the new column in relation *byte_count* are *timestamp* and *total_bytes*. The data values are extracted from *time_dat* a relation which is grouped by *timestamp* and sum of the bytes of grouped data is calculated as *total_bytes*. Each group of data generated in this case are time stamps at which the requests were received at the server.

As shown in Code Snippet 3, we load all the *data* into *ip_data* which is grouped by

ip_add, IP address of the access log input files. Then for each of data loaded in *ip_data*,

ip_counts is generated as the total visits.

Code Snippet 3. ip_data = GROUP data by ip_add; DESCRIBE ip_data; ip_count = FOREACH ip_data GENERATE group AS timestamp, COUNT(data) AS total visits;

The count of the total number of requests is received from a specific IP address and is generated to get the total number of visits by the user. The query is further executed to sort rank the data based on total visits to get the time at which maximum visits are recorded as shown in Code Snippet 4.

```
Code Snippet 4.
    sort_data = RANK ip_count BY total_visits DESC;
    DUMP sort_data;
    STORE ip_count into
    '/user/AccessLogAnalysis/OutPut3' USING PigStorage
    (',');
```

Finally, *DUMP* keyword dumps the result and stores output to the specified output file location using *STORE* command as shown in Code Snippet 4.

The *PigStorage* function used in above Code Snippet 4 is an inbuilt read function with the space separated arguments to read data. Data schemas are declared when column based operations is executed. We define name of each column and column data type. The column data type by default is, *chararray* data type but there are also other built-in data type supported by Pig. *Port* and *bytes* columns are integer data type and other columns take chararray as default.

Log analysis is crucial and contains a lot of information. *DESCRIBE* command can describe any relation. This can be useful when we need to understand how 'Join' and 'Group' statements are used in a particular relation. DESCRIBE time_data describes time_data and using other statements like '*DISTINCT'*, '*FILTER'*, '*GROUP BY*' are useful to eliminate data redundancy.

Filters can be used to filter data on the basis of user requirements. This case study filters all the data based IP addresses which has Port number as 200.

Filtered_data = FILTER ip_data by Port == 200;

The output here is also created in HDFS as shown in Figure 4.

```
grunt> sort_data = Rank ip_count BY numberOfVisits DESC;
grunt> DUMP sort_data;
```

Figure 4. Dump result to generate output.

Result. By analyzing and processing log files in the Pig system, we are able to get the visits of specific user, visit per unit time and failed request. As shown in Figure 4, the number

of views of per web page is retrieved by using the *DUMP* command and the result is printed in the grunt shell without storing them into a file. The analysis is further carried and stored into database to make data available for end users. When the Pig system completes its analysis, the output will be stored in HDFS and a complete report of file location will be available as shown in Figure 5.

| HadoopVersion PigVersion UserId StartedAt FinishedAt 2.7.1 0.16.0 hadoop1 2017-02-08 23:44:02 2017-02-08 23:47:02 | Features GROUP_BY,DISTINCT |
|---|--|
| Success! | |
| Job Stats (time in seconds): JobId Maps Reduces MaxNapTime MinMapTime AvgMapTime job_local825091494_0004 2 1 n/a n/a n/a n/a job_local994152450_0003 2 1 n/a n/a n/a n/a | MedianMapTime MaxReduceTime MinReduceTime AvgReduceTime MedianReducetime Alias Feature Outputs n/a n/a n/a n/a ip_count,ip_data GROUP_BY,COMBINER /user/AccessLogAnalysis/OutPut1, n/a n/a n/a n/a data DISTINCT |
| Input(s): Successfully read 1891715 records (2031283308 bytes) from: "/user/Acces | ssLogAnalysis/WebLogs/AccessLogFile1" |
| Output(s): Successfully stored 81983 records (1607741976 bytes) in: "/user/Accessl | LogAnalysis/OutPut1" |
| Counters: Total records written : 81983 Total bytes written : 1607741976 Spillable Memory Manager spill count : 0 Total bags proactively spilled: 0 Total records proactively spilled: 0 | |
| Job DAG: job_local994152450_0003 -> job_local825091494_0004, job_local825091494_0004 | |

Figure 5. Pig output log.

As shown in Figure 5, after executing the DUMP Command, the output is stored as a output file in Hadoop Distributed File System, HDFS location. The first couple of lines gives the brief summary of the job. Hadoop version, Pig version, UserId, startedAt is the time Pig submits the job, not the time the first job starts running the Hadoop cluster. FinishedAt is the time Pig finishes processing the job, which will be slightly after the time the last MapReduce job finishes.

The section labeled as Job Stats as shown in Figure 5, gives a breakdown of each MapReduce job that was run. They include how many map and reduce tasks each job has, in our case we have one map and one reduce task. Also, include statistics on how long these tasks took and a mapping of aliases in the Pig Latin script to the jobs. The input, output and Counters sections are self-explanatory. In the above Figure 5, total 1891715 records is from the input *AccessLogFile1* and 81983 records is stored in the output file in a structured format.

| URL | Total |
|---------|--------|
| | Visits |
| taltal | 3749 |
| triton | 3736 |
| acm.org | 3712 |
| apt.com | 3672 |
| asi.com | 3650 |
| bix.com | 3597 |
| cml.com | 3579 |
| crl.com | 3561 |
| dfw.net | 3178 |
| diab.se | 3116 |
| efn.org | 3017 |
| flo.org | 2946 |
| fsd.com | 2669 |
| fwi.com | 2597 |
| amsa.ch | 2555 |

Figure 6. Output sample.

As shown in Figure 6, the table shows the lists of URL's and the count of visits per

URL after executing the webServerLog.pig file in the Pig system (on the next page).

The group of operation forces a reduce phase where we can defined the level of parallelism to be implemented. If no parallelism is specified, the number of reducers are

calculated by using formula:

```
reducers = (int)Math.ceil((double)totalInputFileSize / bytesPerReducer)
TotalReducers = Math.min(maxReducers, reducers)
```

Where,

maxReducersNumber is 999 by default

bytesPerReducer is 1073741824 (1GB) bytes by default.

TotalReducers = Min(999, (2031283308/1073741824))

= 1.81978

~ 1 reducers

In the above screenshot on Figure 5, the number of maps are 2 and reducers is 1 and

on input 1891715 records were read whereas on the output 81983 records were reduced and

stored in HDFS location as an output file.

Code (webserverlog.Pig). The code shown below is the Pig Latin script that is used to load *AccessLogFile3* file and defined the our own schema for the input which is space delimited. The ouput is generated and stored in HDFS file location as a *OutPut3* file in a structured format seperated by comma.

```
data = load
'/user/AccessLogAnalysis/WebLogs/AccessLogFile3'
using PigStorage (' ') AS (
ip add: chararray,
temp1: chararray,
temp2: chararray,
timestamp: chararray,
timeZone: chararray,
cs method: chararray,
cs uri: chararray,
request dest: chararray,
port: int,
bytes: int
);
data = DISTINCT data;
time data = GROUP data BY timestamp;
DESCRIBE time data;
byte count = FOREACH time data GENERATE group AS
timestamp, SUM(data.bytes) AS total bytes;
ip data = GROUP data by ip add;
DESCRIBE ip data;
ip count = \overline{FOREACH} ip data GENERATE group AS
timestamp, COUNT(data) AS total visits;
```

```
sort_data = RANK ip_count BY total_visits DESC;
DUMP sort_data;
STORE ip_count into
'/user/AccessLogAnalysis/OutPut3' USING PigStorage
(',');
```

As this code is executed, the jobs runs and we see the status printed in the grunt shell

as number of records generated as output, number of records written, total bytes written, and

job listed Direct Acyclic Graph (DAG).

Performance. The files when used in local mode took approximately 3 minutes to

process each web log files whereas when ran on the MapReduce mode it took approximately

5 minutes to process the same log file of same size as shown in Figure 7 and Figure 8.



Figure 7. Processing time of Pig Script on local.

1817-02-09 21:17:32,446 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - Success! 1817-02-09 21:17:32,528 [main] INFO org.apache.pig.Main - Pig script completed in 5 minutes, 30 seconds and 689 milliseconds (330689 ms) 1adoop1@neetu-VirtualBox:-S

Figure 8. Processing time of Pig Script on MapReduce Mode.

Table 2

Performance Evaluation on Local Mode and MapReduce Mode

| Mode | Time Taken | File Size |
|----------------|--------------------|--------------------|
| Local Mode | 3 minutes (approx) | 283128383308 bytes |
| MapReduce Mode | 5 minutes (approx) | 283128383308 bytes |

Although the processing time in local mode seems more efficient than MapReduce mode as shown in Table 2. The results shown above is generated when the map task is only one (one server), but what happens when our job has 10,000 map tasks? Local mode fails to execute as it runs locally on only one server (single node) therefore when the map task is increased or when data is processed through large clusters (multiple node), MapReduce mode seems to be more efficient than local mode.

In all above processes of inserting unstructured input log files in Apache Pig processing system, implementing and generating some useful information from the logs, we are able to get that in Apache Pig framework using some of our user-defined function. As the input file size increases, each MapReduce stage performs certain optimizations and the job gets reduced drastically. Pig processing is useful to build behavior prediction model based on the user interaction within a website. Using Pig-based flow analyzer the analysis on web server logs files is done with increased convenience and easiness without any in-depth knowledge of programming expertise. Using typical MapReduce paradigm, MapReduce task division module and inbuilt Pig compiles to get the total visits per URL within the given time frame. In the next chapter, we will conclude the output results from the research and propose some future works from our analysis.

Chapter 4: Conclusions and Future Work

With the types of data growing everywhere, the management of data is a great challenge. To manage data, all organizations must develop a data-driven culture, where collection and storage are the integral part of the business. In order to handle this complexity, organizations needs a combined knowledge of analyzing all the types of data that carry some sort of information related to the firm. This paper reveals how these complex weblog files is analyzed using one of the Hadoop data analytics tool, Apache Pig. Pig can easily process data that were extracted from diverse sources and process them using its simple scripting language known as Pig Latin. The data is processed using several Pig Latin statements and filtered out in order to remove data redundancy. The data to processed in this paper is downloaded from NASA web server log files, stored in HDFS, and is accessed around the cluster to perform MapReduce task. From the results achieved, it shows that large amount of unstructured weblogs files will generate useful information in a structured form when used on Apache Pig processing system.

Big data implementation is growing recently and among different types of big data framework, Apache Pig is a work in progress. It is an open-source project and is being actively worked by Yahoo, Facebook, as well as several other health and external contributors to gather useful information from the unstructured data. From year 2006 when Pig was first discovered till recent, Pig itself has developed its own language that includes several Pig Latin statements, data types, general and relational operators, and user-defined functions (UDFs). By implementing Apache Pig, web server log analysis is much simpler and the time consumed for researching on the complex control loops and programming constructs for implementing on a typical local mode, for one map job or MapReduce mode for several maps jobs decreased by a greater percentage.

Beside from processing data, the data that comes in can be in any state and might contain some useful and redundant data as well. The processing system must not only be able to process the data but also support data cleansing and profiling. Apache Pig is enriched with these features to overcome the data quality concerns. Aggregated data from Pig is a minimal subset of data and is loaded to Data Warehouse for Business Analytics and Enterprises. Reporting. Google uses the same algorithm to improve performance by examining the user behavior. Once the data gets inside the Map-Reduce cluster as input, the data goes through several processes like map and reduce and a refined output is generated in Hadoop distributed file system. The data inside Map-Reduce cluster also supports on adding additional data to the clusters without having re-index all over again. Many companies use Hadoop to analyze highlevel queries that were a big challenge a few years ago. The data inside Map-Reduce cluster also generates an iterative processing model and can keep track of every new updates by joining the behavioral model with the user data. This feature is widely used in socialnetworking sites like Facebook and micro-blogging sites like Twitter. Pig cannot be considered as effective as MapReduce when it comes to processing small data or scanning multiple records in random order.

Future work can be extended on using Pig framework to implement unstructured weblog access files on several large clusters for prediction analysis to predict which next page will be visited by the user, implementing some artificial intelligence artifacts. Also, another area where we can extend our future research can be, to examine the MapReduce job when the job has 10, 000 of map tasks instead of one and how efficient will be the system.

References

- Beach, C., & Schiefelbein, W. R. (2013, December 31). Unstructured data: How to implement an early warning system for hidden risks. *Journal of Accountancy*. Retrieved from http://www.journalofaccountancy.com/issues/2014/jan/20126972.html.
- Devakunchari, R. (2014). Handling big data with Hadoop toolkit. *International Conference on Information Communication and Embedded Systems (ICICES2014)*. doi:10.1109/ icices.2014.7033839

Elizabeth, M. (2013). Capturing the value of unstructured data. SAS Institute, Inc.

- Eluri, V. R., Ramesh, M., Al-Jabri, A. S. M., & Jane, M. (2016). A comparative study of various clustering techniques on big data sets using Apache Mahout. 2016 3rd MEC International Conference on Big Data and Smart City (ICBDSC), Muscat, 15-16 March 2016, 1-4. doi.org/10.1109/ICBDSC.2016.7460397.
- Francis, N., & Kurian, K, S. (2015). Data processing for big data applications using Hadoop framework. *IJARCCE*. doi:10.17148/ijarcce.2015.4343
- Gates, A. F., Natkovich, O., Chopra, S., Kamath, P., Narayanamurthy, S. M., Olston, C., ...
 Srivastava, U. (2009). Building a high-level data flow system on top of map-reduce. *Proceedings of the VLDB Endowment*, 2(2), 1414-1425. doi:10.14778/1687553.
 1687568
- Gupta, B., & Kiran, J. (2014). Big data analytics with Hadoop to analyze targeted attacks on enterprise data. *International Journal of Computer Science and Information Technologies (IJCSIT)*, 5(3), 3867-3870.
- Hadoop ecosystem: An introduction (2016). *International Journal of Science and Research* (*IJSR*), 5(6), 557-562. doi:10.21275/v5i6.nov164121

- Herzig, M. D. (2011). Hybrid search ranking for structured and unstructured data. *The Institute AIFB, Karlsruhe Institute of Technology, Germany. Part II, LNCS 6644*, pp. 518-522.
- Holzinger, A., & Pasi, G. (2013). Human-computer interaction and knowledge discovery in complex, unstructured, big data. *Proceedings of the Third International Workshop, HCI-KDD 2013*, SouthCHI 2013, Maribor, Slovenia, July 1-3, 2013. Berlin: Springer.
- Insights on governance, risk and compliance. (2010). Retrieved January 29, 2017, from http://ey.com/GRCinsights.
- Jain, A. (2013). Instant Apache Sqoop. Packt Publishing, Ltd.
- Lai, W. K., Chen, Y.-U., Wu, T.-Y., & Obaidat, M. S. (2013). Towards a framework for large-scale multimedia data storage and processing on Hadoop platform. *The Journal* of Supercomputing, 68(1), 488-507. doi:10.1007/s11227-013-1050-4
- Minsker, M. (2015, January). *Is Hadoop worth the hype*. Retrieved January 29, 2017, from www.destinationCRM.com.
- Nandimath, J., Banerjee, E., Patil, A., Kakade, P., Vaidya, S., & Chaturvedi, D. (2013). Big data analysis using Apache Hadoop. 2013 IEEE 14th International Conference on Information Reuse & Integration (IRI). doi:10.1109/iri.2013.6642536
- NASA-HTTP. (n.d.). Retrieved January 24, 2017, from http://ita.ee.lbl.gov/html/contrib/ NASA-HTTP.html.
- Qazi, R. U. R., & Sher, A. (2016). Big data applications in businesses: An overview. *The International Technology Management Review*, 6(2), 50. doi:10.2991/itmr.2016.6.2.3

Appendix

A. System Specifications and Requirements

- 1. Ubuntu v16.0.2
- 2. Hadoop v2.7.1
- 3. Pig v0.16.0 (r1746530)

B. HDFS Web User Interface

1. Hadoop Distributed File System Overview Page

Apart from command line interface, Hadoop provides web user interface of HDFS resource manager. The HDFS interface is useful in pseudo-distributed mode and fully distributed mode. Image 1, shows the overview of the version number, cluster Id, and block pool Id of the HDFS version used.

| 🗲 🛈 🤁 🛛 localhost:50070/dfshealth.html#tab- | overview | | | | | | |
|---|------------|-----------------|------|---|-----------------------------|--------------|--|
| | Hadoop | Overview | | | | | |
| | Overv | /iew 'lo | |)00' (active) Jan 10 22:04:57 CST 2017 | | | |
| | Version: | | 2.7. | 1, r15ecc87ccf4a0228f35af08f | c56de536e6c | e657a | |
| | Compiled | : | 201 | 5-06-29T06:04Z by jenkins fror | n <mark>(d</mark> etached f | rom 15ecc87) | |
| | Cluster ID |): | CID | 0479cb32-cac9-469e-8fa5-2cf | 1dfbe0b7a | | |
| | Block Poo | I ID: | BP-5 | 564823193-127.0.1.1-14841074 | 485170 | | |

Image 1. HDFS web user interface Overview

2. <u>Node</u>

As shown in image 2, the summary of the configured HDFS UI which is used in our research. It includes the total size available for configuration, data node usages, number of live nodes and dead nodes available, total number of under-replicated blocks, and block deletion start time. The live Node is the node whose children are currently being used during the process.

| Summary | |
|---|-------------------------------|
| Security is off. Safemode is off. 105 files and directories, 49 blocks = 154 total filesystem object(s). Heap Memory used 43.62 MB of 65.88 MB Heap Memory. Max Heap Memory is 966.69 MB. Non Heap Memory used 60.84 MB of 61.84 MB Commited Non Heap Memory. Max Non Heap Memory | is -1 B. |
| Configured Capacity: | 94.19 GB |
| DFS Used: | 286.67 MB (0.3%) |
| Non DFS Used: | 13.04 GB |
| DFS Remaining: | 80.87 GB (85.85%) |
| Block Pool Used: | 286.67 MB (0.3%) |
| DataNodes usages% (Min/Median/Max/stdDev): | 0.30% / 0.30% / 0.30% / 0.00% |
| Live Nodes | 1 (Decommissioned: 0) |
| Dead Nodes | 0 (Decommissioned: 0) |
| Decommissioning Nodes | 0 |
| Total Datanode Volume Failures | 0 (0 B) |
| Number of Under-Replicated Blocks | 40 |
| Number of Blocks Pending Deletion | 0 |
| Block Deletion Start Time | 1/10/2017, 10:04:57 PM |

Image 2. HDFS web user interface showing summary of configuration

3. Data Node Information

A DataNode stores data in HDFS. A functional file system may have more than one DataNode, with data replicated across them. The DataNode information shown in Image 3 below shows the given DataNode available in our HDFS system. "neetu-VirtualBox:50010", is in the service state with capacity of 94.19GB and 49 Blocks. DataNode further performs read-write operations on the file system, as per client request.

| Datanode Inf | ormatio | n | | | | | | | | |
|---|-----------------|----------------|----------|--------------|-----------------|-----------|--------|---------------------|-------------------|---------|
| n operation | | | | | | | | | | |
| Node | Last contact | Admin State | Capacity | Used | Non DFS Used | Remaining | Blocks | Block pool used | Failed Volumes | Version |
| neetu-VirtualBox:50010 (127.0.0.1:50010) | 0 | In Service | 94.19 GB | 286.67 MB | 13.04 GB | 80.87 GB | 49 | 286.67 MB (0.3%) | 0 | 2.7.1 |

Image 3. HDFS web user interface showing DataNode Information

4. NameNode Information

The NameNode is the centerpiece of an HDFS. It allows user to browse the file system. NameNode acts as a master server and manages the file system namespace. NameNode also provides some logs information about current insight situation representing the transaction Id, journal manager location, and state. The edit log files contains the log records of the current situation. The NameNode directory gets configured manually on hdfs-site.xml while installing Hadoop.

| NameNode Journal Stat | tus | |
|---|---|--------|
| Current transaction ID: 1075 | | |
| Journal Manager | State | |
| FileJournalManager(root=/tmp/hadoop-hadoop1 /dfs/name) | EditLogFileOutputStream(/tmp/hadoop-hadoop1/dfs/name/current /edits_inprogress_000000000000001075) | |
| NameNode Storage | | |
| Storage Directory | Туре | State |
| /tmp/hadoop-hadoop1/dfs/name | IMAGE_AND_EDITS | Active |
| Hadoop, 2015. | | |

Image 4. HDFS web user interface showing NameNode Journal Status and Storage

5. Browse directory

The image 5 shown below is the directory shown on HDFS file system. The UI has several sections, Overview, Datanodes, Snapshots, startup progress, and utilities. The browser directory listed below shows the directory created inside HDFS to store the input files and store output results. Different level of permissions can be assigned to the files and owner can be specified per each files.

| | Browse | Director | у | | | | | | | |
|----------------|--------------------------------------|--------------|--------------------------|-------|---------------------------------------|-------------|------------|---------|---------|---|
| | 1 | | | | | | | | | G |
| | Permission | Owner | Group | Size | Last Modified | Replic | ation Blo | ck Size | Name | |
| | drwxr-xr-x | hadoop1 | supergroup | 0 B | 1/16/2017, 8:52:10 PM | 0 | 0 B | | tmp | |
| | drwxr-xr-x | hadoop1 | supergroup | 0 B | 1/21/2017, 12:57:15 PM | 0 | 0 B | | user | |
| Brows | Hadoop, 2015. e Direct | ory | | | | | | | | |
| | e Direct | • | | | | | | | | |
| /user/AccessLo | e Direct | ngs | Size | lat | Modified | Peolication | Block Size | Name | | |
| | e Direct | • | Size 195.73 MB | | : Modified 2017, 8:53:07 PM | Replication | Block Size | Name | ogFile1 | |
| /user/AccessLo | e Direct gAnalysis/WebLc owner | ogs Group | | 2/8/2 | | | | | - | |

Image 5. HDFS web user interface Directory

C. MapReduce Processes

1. Logs Details

As shown in image 6 below, the log shows which files is being executed from

which location. In our case, as shown in image 6, we are executing our file from

HDFS location. The log shows the name of the running Job (job_145478739_8002),

saved output task location, status of the reduce task execution, etc.

| 2017-02-08 22:54:32,890 [LocalJobRunner Map Task Executor #0] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.PigRecordReader - Curren |
|---|
| t split being processed hdfs://localhost:9000/tmp/temp-1557744124/tmp2107943726/part-r-00000:0+134217728 |
| 2017-02-08 22:54:32,956 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - (EQUATOR) 0 kvi 26214396(104857584) |
| 2017-02-08 22:54:32,956 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - mapreduce.task.io.sort.mb: 100 |
| 2017-02-08 22:54:32,956 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - soft limit at 83886080 |
| 2017-02-08 22:54:32,956 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - bufstart = 0; bufvoid = 104857600 |
| 2017-02-08 22:54:32,956 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - kvstart = 26214396; length = 6553600 |
| 2017-02-08 22:54:32,972 [LocalJobRunner Map Task Executor #0] INFO org.apache.hadoop.mapred.MapTask - Map output collector class = org.apache.hadoop.map |
| red.MapTaskSMapOutputBuffer |
| 2017-02-08 22:54:32,985 [LocalJobRunner Map Task Executor #0] INFO org.apache.pig.impl.util.SpillableMemoryManager - Selected heap (Tenured Gen) of size |
| 699072512 to monitor. collectionUsageThreshold = 489350752, usageThreshold = 489350752 |
| 2017-02-08 22:54:32,988 [LocalJobRunner Map Task Executor #0] WARN org.apache.pig.data.SchemaTupleBackend - SchemaTupleBackend has already been initiali |
| |
| 2017-02-08 22:54:59,510 [pool-9-thread-1] WARN org.apache.pig.data.SchenaTupleBackend - SchenaTupleBackend has already been initialized |
| 2017-02-08 22:54:59,532 [pool-9-thread-1] INFO orq.apache.pip.backend.hadoop.executionengine.mapReducelayer.PigHapReduceSReduce - Allases being processed per job phase (AllasName[line.off |
| set]): M: byte count[6,13],time data[4,12] C: byte count[6,13],time data[4,12] R: byte count[6,13] |
| 2017-02-08 22:55:04,164 [communication thread] INFO org.apache.hadoop.mapred.LocalJobRunner - reduce > reduce > |
| 2017-02-08 22:55:04,206 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.NapReduceLauncher - 98% complete |
| 2017-02-08 22:55:04,206 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.HapReduceLauncher - Running jobs are [job_local145478739_0002] |
| 2017-02-08 22:55:05,155 [pool-9-thread-1] INFO org.apache.hadoop.mapred.Task - Task:attempt_local145478739_0002_r_000000_0 is done. And is in the process of committing |
| 2017-02-08 22:55:05,160 [pool-9-thread-1] INFO org.apache.hadoop.mapred.LocalJobRunner - reduce > reduce |
| 2017-02-08 22:55:05,160 [pool-9-thread-1] INFO org.apache.hadoop.mapred.Task - Task attempt local145478739_0002_r_0000060 is allowed to commit now |
| 2017-02-08 22:55:05 [pool-9-thread-1] INFO org.apache.hadop.napreduce.lib.output.FileOutputConnitter - Saved output of task 'attempt_local145478739_0002_r_000000_0' to hdfs://localho |
| st:9000/tmp/temp-1557744124/tmp974198704/_temporary/0/task_local145478739_0002_r_000000 2017-02-08 22:55:05.188 [pool-9-thread-1] INFO_org.apache.hadoop.mapred.LocalJobRunner - reduce > reduce |
| 2017-02-08 22:55:65,192 [Doul-9-thread-1] INFO org.apache.hadoop.napred.tocka.cuoka.unudamieri - resulter 2 resulter 3 doule - 000000 0' done. |
| 2017-02-08 2255:05.196 [pool-9-timed-1] INTO org-apache-indoop.mapred.ioc3106kmner - Finishing task: attempt local149478739 0002 [000009 0 00006. |
| 2017-02-08 22:55:06,197 [https://jinko.jadop.maprel.cal.bokumer - reduce task secutor complete. |
| 2017-02-08 22:55:07,713 [main] INFO org.apache.hadoop.metrics.jvm.JvmMetrics - Cannot initialize JVM Metrics with processName=JobTracker, sessionId= - already initialized |
| 2017-02-08 22:55:07,717 [main] INFO org.apache.hadoop.metrics.jvn.JvnMetrics - Cannot initialize JVM Metrics with processName=JobTracker, sessionId= - already initialized |
| 2017-02-08 22:55:07,718 [main] INFO org.apache.hadoop.metrics.jvn.JvmWetrics - Cannot initialize JVM Metrics with processName=JobTracker, sessionId= - already initialized |
| 2017-02-08 22:55:07,783 [main] INFO org.apache.pig.backend.hadoop.executionengine.mapReduceLayer.MapReduceLauncher - 100% complete |
| 2017-02-08 22:55:07,804 [main] INFO org.apache.pig.tools.pigstats.mapreduce.SimplePigStats - Script Statistics: |

Image 6. Logs while executing input file in Pig Grunt shell

2. Output log Report

As shown in image 7, the log reports:

- I. Successfully read 1891715 records
- II. Successfully stored records 1099799 files.



Image 7. Output log report in Pig Grunt shell D.

D. Input log file sample sampshot

| File Edit Format View Help 199.72.81.55 - [01/Jul/1995:00:00:01 - 0400] "GET /history/apollo/ HTTP/1.0" 200 6245unicomp6.unicomp.net - [01/Jul/1995:00:00:06 - 0400] "GET /shuttle/courdown/count.gif HTTP/1.0" 200 40310unicomp6.unicomp.net - [01/Jul/1995:00:00:04 - 0400] "GET /shuttle/courdown/count.gif HTTP/1.0" 200 1473205.189.154.54 - [01/Jul/1995:00:00:24 - 0400] "GET /image/launchpalma-small.gif HTTP/1.0" 200 11473205.189.154.54 - [01/Jul/1995:00:00:24 - 0400] "GET /shuttle/coundown/HTTP/1.0" 200 12054cheyer.clark.net - [01/Jul/1995:00:00:38 - 0400] "GET /shuttle/missions/sts-71/movies/sts-71/movies/sts-71/mine-dock-2.mgp HTTP/1.0" 200 398! HTTP/1.0" 200 12054cheyer.clark.net - [01/Jul/1995:00:00:38 - 0400] "GET /shuttle/missions/sts-71/movies/sts-71/movies/sts-71/mines/sto.71/moves/sto.71/mines/sto.71/moves/sto.71/moves/sto.71/moves/sto.71/mines/sto.71/moves/sto.71/mines/sto.71/moves/sto.71/mines/sto.77/mines/sto.71/mines/sto.71/mines/sto. | 199.72.81.55 - [01/Jul/1995:00:00:01 -0400] "GET /history/apollo/ HTTP/1.0" 200 6245unicomp6.unicomp.net - [01/Jul/1995:00:00:14 -0400] "GET /shuttle/courtdown/court.gif HTTP/1.0" 200 40310unicomp6.unicomp.net - [01/Jul/1995:00:00:14 -0400] "GET /shuttle/courtdown/court.gif HTTP/1.0" 200 40310unicomp6.unicomp.net - [01/Jul/1995:00:00:24 -0400] "GET /image asahi-net.or.jp - [01/Jul/1995:00:00:24 -0400] "GET /image/sole:02:40 -0400] "GET /image/sole:02:40 -0400] "GET /image/sole:02:40 -0400] "GET /image/sole:02:400] "GET /image/sole:02:400] "GET /image/sole:02:400] "GET /image/sole:02:400] "GET /shuttle/countdown/ HTTP/1.0" 200 308:400] "GET /shuttle/missions/sts-71/images/images.html HTTP/1.0" 200 7634ww-al.proxy.aol.com - [01/Jul/1995:00:00:34 -0400] "GET /shuttle/missions/sts-71/images/images.html HTTP/1.0" 200 7634ww-al.proxy.aol.com - [01/Jul/1995:00:00:35 -0400] "GET /shuttle/missions/sts-71/images/images.html HTTP/1.0" 200 7634ww-al.proxy.aol.com - [01/Jul/1995:00:00:35 -0400] "GET /shuttle/missions/sts-71/images/images.html HTTP/1.0" 200 7634ww-al.proxy.aol.com - [01/Jul/1995:00:00:35 -0400] "GET /shuttle/resources/orbiters/endeavour.gif HTTP/1.0" 200 1520pp-nc/200 152 -0400] "GET /shuttle/resources/orbiters/endeavour.gif HTTP/1.0" 200 7674ix/sorbite/sorbiters/endeavour.gif HTTP/1.0" 200 7674ix/sorbite/sorbiters/endeavour.gif HTTP/1.0" 200 7674ix/sorbite/sorbiters/endeavour.gif HTTP/1.0" 200 7674ix/sorbie-06.ix.netcom.com - [01/Jul/1995:00:01:52 -0400] "GET /shuttle/courtdown/vise/sits-78/mission-sts-78.html HTTP/1.0" 200 7674ix/sorbie-06.ix.netcom.com - [01/Jul/1995:00:01:52 -0400] "GET /shuttle/courtdown/vise/sits-78/mission-sts-78.html HTTP/1.0" 200 7674ix/sorbie-06.ix.netcom.com - [01/Jul/1995:00:02:16 -0400] "GET /shuttle/courtdown/vise/sits-75/mission-sts-78.html HTTP/1.0" 200 7674ix/sorbie-06.ix.netcom.com - [01/Jul/1995:00:02:16 -0400] "GET /shuttle/courtdown/HTTP/1.0" 200 3085d; T /shuttle/courtdown/vise/sits-55/atc/100:01:20 +0400] "GET /shuttle/courtdown/HTTP/1.0" 200 7674ix/sorbie-05:15 -50+20 | 🔄 AccessLogFile1 - Notepad | 3 |
|--|--|--|---|
| <pre>p.net [01/Jul/1995:00:00:14 -0400] "GET /shuttle/countdown/count.gif HTTP/1.0" 200 40310unicomp5. unicomp.net [01/Jul/1995:00:00:14 -0400] "GET /images/lauchhaplms-small.gif HTTP/1.0" 200 11473205.189.154.54 [01/Jul/1995:00:00:24 -0400] "GET /images/NASA-logosmall.gif HTTP/1.0" 200 786ix-orl2-01.ix.netcom.com [01/Jul/1995:00:00:41 -0400] "GET /shuttle/countdown/HTTP/1.0" 200 49152ppp-nyc-i 05.189.154.54 [01/Jul/1995:00:00:01:08 -0400] "GET /shuttle/missions/sts-71/movies/sts-71-mir-dock-2.mgp HTTP/1.0" 200 49152ppp-nyc-i 05.189.154.54 [01/Jul/1995:00:01:08 -0400] "GET /shuttle/missions/sts-71/movies/sts-71-mir-dock-2.mgp HTTP/1.0" 200 7340ww-al.proxy.aol.com [01/Jul/1995:00:01:19 -0400] "GET /shuttle/missions/sts-71/images/Kages/te-04: istory/apollo/images/footprint-small.gif HTTP/1.0" 200 1814205.189.154.54 [01/Jul/1995:00:01:19 -0400] "GET /shuttle/missions/sts-71/images/KGC-95EC-04: mages/lauchhaudowing if HTTP/1.0" 200 70741ink097.txdirect.net [01/Jul/1995:00:01:44 -0400] "GET /shuttle/missions/sts-78/mission-sts-78.html HTTP/1.0" 200 7051095:00:01:57 -0400] "GET / HTTP/1.0" 200 70741ink097.txdirect.net [01/Jul/1995:00:01:44 -0400] "GET /shuttle/missions/sts-78/mission-sts-78.html HTTP/1.0" 200 7061/Jul/1995:00:01:52 -0400] "GET /images/MASA-logosmall.gif HTTP/1.0" 200 786199.72.81.55 [01/Jul/1995:00:01:52 -0400] "GET /images/MASA-logosmall.gif HTTP/1.0" 200 70741ink097.txdirect.net [01/Jul/1995:00:02:15 -0400] "GET /shuttle/missions/sts-78/mission-sts-78.html HTTP/1.0" 200 70741ink097.txdirect.net [01/Jul/1995:00:02:15 -0400] "GET /shuttle/countdown/life/file/file/file/file/file/file/file</pre> | <pre>p.net [01/JU/1995:00:00:14 -0400] "GET /insges/laurchpains-small.gif HTTP/1.0" 200 40310uricomp.net [01/JU/1995:00:00:14 -0400] "GET /insges/star.pages/laurchpains-small.gif HTTP/1.0" 200 12051ches/star.pages/laurchpains-small.gif HTTP/1.0" 200 12051ches/star.pages/laurchpains-small.gif HTTP/1.0" 200 12051ches/star.pages/laurchpains-small.gif HTTP/1.0" 200 12051ches/star.pages/laurchpains-small.gif HTTP/1.0" 200 13051ches/star.pages/laurchpains-small.gif HTTP/1.0" 200 40152pp-nyc-i 05.189.154.54 [01/JU/1995:00:01:19 -0400] "GET /intutle/missions/star.pi/meins/star.pi</pre> | File Edit Format View Help | |
| 995:00:04:05 -0400] "GET /images/MOSAIC-logosmall.gif HTTP/1.0" 200 363129.188.154.200 [01/Jul/1995:00:04:05 -0400] "GET /images/USA-logosmall.gif HTTP/: apollo/images/apollo-logo.gif HTTP/1.0" 200 3047unicomp6.unicomp.net [01/Jul/1995:00:04:16 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.154.54 [01/Jul/1995:00:04:16 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.154.54 [01/Jul/1995:00:04:25 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.154.54 [01/Jul/1995:00:04:25 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.154.54 [01/Jul/1995:00:04:25 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.154.54 - [01/Jul/1995:00:04:25 -0400] "GET /ksc.html HTTP/1.0" 200 707425.189.571/movies/sts-71/mates/KSC-logosmall.gif HTTP/1.0" 200 707425.199.580:04:44 -0400] "GET /ksc.html HTTP/1.0" 200 707425.199.580:04:44 -0400] "GET /ksc.html HTTP/1.0" 200 707425.199.580:04:25 -0400 "GET /ksc.html HTTP/1.0" 200 707425.199.580:04:44 -0400] "GET /ksc.html HTTP/1.0" 200 707425.199.580:04:44 -0400] "GET /ksc.html HTTP/1.0" 200 707405.34 -0400] "GET /ksc.htm | -71/images/images.html HTTP/1.0" 200 7634ix-sd11-26.ix.netcom.com - [01/Jul/1995:00:05:06 -0400] "GET /cgi-bin/imagemap/countdown?107,144 HTTP/1.0" 302 96: m.com - [01/Jul/1995:00:05:17 -0400] "GET /icons/menu.xbm HTTP/1.0" 200 527net-1-141.eden.com - [01/Jul/1995:00:05:17 -0400] "GET /shuttle/missions/sts-; ect.net - [01/Jul/1995:00:05:23 -0400] "GET /icons/menu.xbm HTTP/1.0" 200 527ix-war-mi1-20.ix.netcom.com - [01/Jul/1995:00:05:23 -0400] "GET /icons/text. | <pre>Fdb Edh Format Vew Help 199.72.81.55 - [01/Jul/1995:00:00:01 -0400] "GET /history/apollo/ HTTP/1.0" 200 6245unicomp6.unicomp.net - [01/Jul/1995:00:00:24 -0400] "GET /images/ asahi-net.or.jp - [01/Jul/1995:00:00:19 -0400] "GET /images/launchpalms-small.gif HTTP/1.0" 200 1473205.189.154.54 - [01/Jul/1995:00:00:24 -0400] "GET / -0400] "GET /images/NASA-logosmall.gif HTTP/1.0" 200 783/sorol-201.ix.netcom.com - [01/Jul/1995:00:00:24 - 0400] "GET /shuttle/countdow/HTTP/1.0" 200 393 HTTP/1.0" 200 12854scheyer.clark.net - [01/Jul/1995:00:00:24 -0400] "GET /shuttle/missions/sts-71/movies/sts-71-mir-dock-2.mge HTTP/1.0" 200 49152ppp-nyc: istory/apollo/images/footprint-small.gif HTTP/1.0" 200 181531jp.yab.com - [01/Jul/1995:00:01:19 -0400] "GET /shuttle/missions/sts-71/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason/sts-73/images/mason</pre> | ^ |

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| File Edit Format View Help | |
| File Edit Format View Help ect.net - [01/Jul/1995:00:05:23 -0400] "GET /icons/menu.xbm HTTP/1.0" 200 527ix-war-mi1-20.ix.netcom.com - [01/Jul/1995:00:05:23 -0400] "GET /icons/ 52 - [01/Jul/1995:00:05:35 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 1204port2.electrotex.com - [01/Jul/1995:00:05:37 -0400] "GET /icons/ ab.ca - [01/Jul/1995:00:05:35 -0400] "GET /ihuttle/missions/sts-71/images/KSC-95EC-0876.gif HTTP/1.0" 200 53398ix-war-mi1-20.ix.netcom.com - [01/Jul/1995:00:05:37 -0400] "GET /shuttle/ .ab.ca - [01/Jul/1995:00:05:35 -0400] "GET /shuttle/missions/sts-71/images/KSC-95EC-0876.gif HTTP/1.0" 200 7074pmd_23.digital.net - [01/Jul/1995:00:05:37 -0400] "GET /shuttle/missions/sts-71/stmages/KSC-95EC-0876.gif HTTP/1.0" 200 7074pmd_23.digital.net - [01/Jul/1995:00:06:03 -0400] "GET /shuttle/missions/sts-71.ptch-small. 7195:00:06:11 -0400] "GET /images/faq.gif HTTP/1.0" 200 263burger.letters.com - [01/Jul/1995:00:06:12 -0400] "GET /shuttle/countdown/liftoff.html HTT ssions/missions.html HTTP/1.0" 200 8677gater4.sematech.org - [01/Jul/1995:00:06:26 -0400] "GET /cgi-bin/imagemap/countdown?88,208 HTTP/1.0" 200 2537ix-wdg-va2-26.ix.netcom.com - [01/Jul/1995:00:06:33 -0400] "GET /initory/apollo-11/images/ HTTP/1.0" 200 538bw-al.proxy.aol.com - [01/Jul/1995:00:06:26 -0400] "GET /initory/apollo/apollo-11/images/HTTP/1.0" 200 538bw-al.proxy.aol.com - [01/Jul/1995:00:06:40 -0400] "GET /images/Mattnew.gif HTTP/1.0" 200 65181jn32.indirect.com - [01/Jul/1995:00:06:41 -04 0 3985ww-al.proxy.aol.com - [01/Jul/1995:00:06:49 -0400] "GET /images/Wattnew.gif HTTP/1.0" 200 65181jn32.indirect.com - [01/Jul/1995:00:06:41 -04 0 3985ww-al.proxy.aol.com - [01/Jul/1995:00:06:49 -0400] "GET /images/Wattnew.gif HTTP/1.0" 200 65181jn32.indirect.com - [01/Jul/1995:00:06:64 -0400] - [01/Jul/1995:00:06:49 -0400] "GET /images/WORLD-logosmall.gif HTTP/1.0" 200 658bhrandt.xensei.com - [01/Jul/1995:00:06:54 -0400] "GET /images/WORLD-logosmall.gif HTTP/1.0" 200 658bhrandt.sensei.com - [01/Jul/1995:00:07:08 -0400] "GET /inttle/ - [01 | text. ^ /cour 1/19: 05:55 .gif P/1.(te27. 0 34! 400] /mis: e/mi: 1995: |
| <pre>ruct.gif HTTP/1.0" 200 1414asp.erinet.com [01/Jul/1995:00:07:27 -0400] "GET /software/winvn/bluemarb.gif HTTP/1.0" 200 4441ppp236.iadfw.net [01/Jul/1995:00:07:34 -0400] "GET /software/winvn/bluemarb.gif HTTP/1.0" 200 4441ppp236.iadfw.net [01/Jul/1995:00:07:37 -0400] "GET /software/winvn/bluemarb.gif HTTP/1.0" 200 9385199.166 - [01/Jul/1995:00:07:37 -0400] "GET /software/winvn/bluemarb.gif HTTP/1.0" 200 9385199.166 - [01/Jul/1995:00:07:37 -0400] "GET /software/winvn/bluemarb.gif HTTP/1.0" 200 1204pm4_23.digital.net [01/Jul/1995:00:07:34 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 1204pm4_23.digital.net [01/Jul/1995:00:07:43 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 1204pm4_23.att.exas.net - s [01/Jul/1995:00:07:43 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 1204dnet018.st.texas.net - [01/Jul/1995:00:07:43 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 1204dnet018.st.texas.net [01/Jul/1995:00:07:43 -0400] "GET /images/KSC-logosmall.gif HTTP/1.0" 200 3985brandt.xensei.com - [01/Jul/1995:00:08:13 -0400] "GET /images/NASA-logosmall.gif HTTP/1.0" 304 0net-1-141.eden.com - [01</pre> | d11-: "GET ET /: ogosr :00:(ul/1! |
| 1.0" 200 786ttyu0.tyrell.net - [01/Jul/1995:00:08:27 -0400] "GET /images/MOSAIC-logosmall.gif HTTP/1.0" 200 363199.120.91.6 - [01/Jul/1995:00:08:28 - (01/Jul/1995:00:08:28 - (01/Jul/1995:00:08:25 - (04/D) "GET /images/MSAIC-logosmall.gif HTTP/1.0" 200 363199.120.91.6 - [01/Jul/1995:00:08:25 - (04/D) "GET /images/MSAIC-logosmall.gif HTTP/1.0" 200 363119.120.91.6 - [01/Jul/1995:00:08:25 - (04/D) "GET /images/MSAIC-logosmall.gif HTTP/1.0" 200 363119.120.91.6 - [01/Jul/1995:00:08:25 - 0400] "GET /images/MSAIC-logosmall.gif HTTP/1.0" 200 363119.120.01.6 - [01/Jul/1995:00:08:25 - 0400] "GET /images/MSAIC-logosmall.gif HTTP/1.0" 200 363119.130.com - [01/Jul/1995:00:08:20 - 0400] "GET /imatory/skylab/skylab-4.html HTTP/1.0" 200 1393pipe6.nyc.pipeline / (11/Jul/1995:00:08:17 - 0400] "GET /imatory/skylab/skylab-4.html HTTP/1.0" 200 1393pipe6.nyc.pipeline / 200 2044166.79.67.111 - [01/Jul/1995:00:09:17 - 0400] "GET /isotry/skylab/skylab-4.html HTTP/1.0" 200 43030.wmproxy.info.au - [01/Jul/1995:00:08:29:17 -0400] "GET /isotry/skylab/skylab-4.html HTTP/1.0" 200 43030.wmproxy.info.au - [01/Jul/1995:00:08:17 -0400] "GET /isotry/skylab/skylab-4.html HTTP/1.0" 200 43030.wmproxy.info.au - [01/Jul/1995:00:09:17 -0400] "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 43030.wmproxy.info.au - [01/Jul/1995:00:09:17 -0400] "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 304 0400] "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 304 0400] "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 304 0401 "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 304 0400] "GET /isotry/skylab/skylab-4.html HTP/1.0" 200 304 0400] "GET /isotry/skylab/skylab/skylab-4.html HTP/1.0" 200 304 0400] "GET /isotry/skylab | tml F ASA-: .com :00:(1204 "GET 0.sp(|
| /icons/blank.xbm HTTP/1.0" 200 509129.188.154.200 - [01/Jul/1995:00:10:37 -0400] "GET /icons/menu.xbm HTTP/1.0" 200 527129.188.154.200 - [01/Jul/1995 -0400] "GET /ksc.html HTTP/1.0" 200 7074129.188.154.200 - [01/Jul/1995:00:10:23 -0400] "GET /shuttle/missions/sts-71/ HTTP/1.0" 200 3373ix-dfw13-20.ix. :31 -0400] "GET /shuttle/missions/S1-1/images/S0HC159.GET HTTP/1.0" 200 78255www-d3.proxy.aol.com - [01/Jul/1995:04:01:23 -0400] "GET /shuttle/missions/sts-71/images/images.html HTTP/1.0" 200 7634brar 200 689129.188.154.200 - [01/Jul/1995:00:10:52 -0400] "GET /icons/image.xbm HTTP/1.0" 200 509teleman.pr.mcs.net - [01/Jul/1995:04:01:33 -0400] "GET /icons/image.xbm HTTP/1.0" 200 509teleman.pr.mcs.net - [01/Jul/1995:04:01:35 -0400] "GET /icons/image.xbm HTTP/1.0" 200 786dynip38.efn.org - [01/Jul/1995:04:01:1:03 -0400] "GET /icons/image.xbm HTTP/1.0" 200 3723115 -0400] "GET /icons/image.xbm HTTP/1.0" 200 37231115 -0400] "GET /icons/image.xbm HTTP/1.0" 200 37231115 -0400] "GET /shuttle/mission r.html HTTP/1.0" 200 37231114.0400 "GET /http://dtmain.pr.mcs.net - [01/Jul/1995:04:11:24 -0400] "GET /shuttle/countdown/count.gif HTTP/1.0" 200 40310teleman.pr.mcs.net - [01/ | .net s/st: ndt.> /shu1 Ju1/: ns/ F |
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D. ukr - [@1/Jul/1995:00:18:13 -0400] "GET /history/apollo/images/footprint-logo.gif HTTP/1.0" 200 4209slip-5.io.com - [@1/Jul/1995:00:18:15 -0400] "GET //i."
/1.0" 200 509www-al.proxy.aol.com - [@1/Jul/1995:00:18:27 -0400] "GET //acilities/vab.html HTTP/1.0" 200 445teleman.pr.mcs.net - [@1/Jul/195:00:18:34 -0400] "GET //sbuttle/technol
m - [@1/Jul/1995:00:18:33 -0400] "GET /history/apollo/as-201/sounds/ HTTP/1.0" 200 372ottgate2.bnr.ca - [@1/Jul/1995:00:18:34 -0400] "GET /sbuttle/technol
ges/lamach-logo.gif HTTP/1.0" 200 1736pbS.earthlight.co.nz - [@1/Jul/1995:00:18:46 -0400] "GET /huttle/missions/sts-77/mages/HTTP/1.0" 200 966cu-dial
untdown/ HTTP/1.0" 200 20845h-shining.norfolk.infi.net - [@1/Jul/1995:00:19:04 -0400] "GET /shuttle/technol
ges/lamages.html HTTP/1.0" 200 736dialup61.afn.org - [@1/Jul/1995:00:19:04 -0400] "GET /shuttle/missions/sts-71/movies.html HTTP/1.0" 200 966cu-dial
untdown/ HTTP/1.0" 200 736dialup61.afn.org - [@1/Jul/1995:00:19:10 -0400] "GET /shuttle/missions/sts-71/movies.html HTTP/1.0" 200 4045y
ges/lamges.html HTTP/1.0" 200 736dialup61.afn.org - [@1/Jul/1995:00:19:10 -0400] "GET /shuttle/missions/sts-71/movies.html HTTP/1.0" 200 4179
ges/lamges.html HTTP/1.0" 200 4530tigtate2.bnr.ca - [@1/Jul/1995:00:19:12 -0400] "GET /shuttle/missions/sts-71/movies.html HTTP/1.0" 200 7930tppl60.iad/m.
.com - [@1/Jul/1995:00:19:35 -0400] "GET /shuttle/canology/mages/et-intertank_i-manll.gif HTTP/1.0" 200 7930tppl60.iad/m.
.com - [@1/Jul/1995:00:19:35 -0400] "GET /shuttle/canology/mages/et-intertank_i-manll.gif HTTP/1.0" 200 7930tppl60.iad/m.
.com - [@1/Jul/1995:00:19:35 -0400] "GET /shuttle/canology/mages/et-intertank_i-manll.gif HTTP/1.0" 200 7930tppl60.iad/m.
.com - [@1/Jul/1995:00:19:35 -0400] "GET /shuttle/missions/sts-71/mission-sts-71.html HTTP/1.0" 200 1236dal31.pic.net - [@1/Jul/1995:00:128:4 -0400] "GET /intert/missions/sts-71/mission-sts-71.html HTTP/1.0" 200 123202.70.0.6.6.5 - [@1/Jul/1995:00:28:12 -0400] "GET /shuttle/missions/sts-63/sounds/ HTTP/1.0" 200 133 [03/02/1995:00:20:36 -0400] GET /shuttle/missions/sts-70/mission-sts-70.html HTTP/1.0* 200 13469ww-ai.proxy.aol.com - [01/10/1995:00:20:36 1995:00:20:44 -0400] "GET /shuttle/technology/images/et_1.jpg HTTP/1.0* 200 14414202.70.0.6 - [01/10/1/1995:00:20:45 -0400] "GET /shuttle/missions/sts-63/r -0918.jpg HTTP/1.0* 200 46888810;n-5.io.com - [01/10/1995:00:22:55 -0400] "GET /icons/sound.xbm HTTP/1.0* 200 330ix-phs5-1.71.x.netcom.com - [01/10/1/1995:00:20:45 -0400] "GET /amages/mages.html HTTP/1.0* 200 7634134.20:175.12 - [01/10/1/1995:00:21:11 -0400] "GET /images/MOSAIC-logosmall.gif HTTP/1.0* 200 363ix-tsml-26.ix.netcor /sts-71/images/images.html HTTP/1.0* 200 7634134.20:175.12 - [01/10/1/1995:00:21:23 -0400] "GET /cgi-bin/imagemap/countdown?110,112 HTTP/1.0* 302 111ix-ftw-ll.gif HTTP/1.0* 200 2124blv-pm2-ip16.haltoyn.com - [01/10/1/1995:00:21:23 -0400] "GET /cgi-bin/imagemap/countdown?110,112 HTTP/1.0* 302 11457 72204.97.234.46 - [01/10/1/1995:00:21:38 -0400] "GET / HTTP/1.0* 200 7074miccom.com - [01/10/1/1995:00:21:39 -0400] "GET /history/apollo/apollo-13/sounds/i 363annex12-36.dial.umd.edu - [01/10/1/1995:00:21:45 -04400] "GET /shuttle/missions/sts-71/images/Ksc-05E0-6350.jpg HTTP/1.0* 200 64427ix-dfw11-01.ix.netcom.ccm - [01/10/1/1995:00:21:50 -04400] "GET /images/USA-logosmall.gif HTTP/1.0* 200 334ix-phx5-17.ix.netcom.ccm - [01/10/1/195:00:21:50 -0440] "GET -71/sts-71/patch-small.gif HTTP/1.0* 200 3635ix-dfw11-01.ix.netcom.ccm - [01/10/1/195:00:21:50 -0440] "GET /shuttle/mission/sts-71/mission-sts-70.htm! HTTP/1.0* 200 40310sam-slip-16.neosoft.ccm - [01/10/1/195:00:21:50 -0440] "GET /shuttle/mission/sts-71/images/MAA-logosmall.gif HTTP/1.0* 200 366suts5p6.cc.ukans.edu - [01/10/1/195:00:22:80 -04400] "GET /shuttle/mission/sts-71/mission-sts-71.htm! HTTP/1.0* 200 40310sam-slip-16.neosoft.com - [01/10/1/195:00:22:13 -0400] "GET /images/MAA-logosmall.gif HTTP/1.0* 200 736sam-slip-16.neosoft.com - [01/20/1/195:00:22:03 -04400] "GET /images/MAA-logosmall.gif HTTP/1.0* 200 736sam-slip-16.neosoft.com - [01/20/

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E. Output sample snapshot

| | untu (02122017) [Running] - Oracle VM VirtualBox |
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| | en.com,18 |
| 1 | fcs.nl,7 |
| 3 | fml.ch,17 |
| | gdc.ca,1 hel.fl,48 |
| 100 | ksm.my.47 |
| | ktt.f1,31 |
| | lvb.nl,1 |
| 124 | pb.net,10 rhh.dk,114 |
| 1000 | sea.ru,18 |
| - | www.sg,26 |
| | RTD.COM,1 |
| | UNAV.ES.8 |
| ~ | aod.com,5 axs.net,3 |
| 1 | bix.com,17 |
| | cml.com,55 |
| 100 | crl.com,45 |
| A | crx.com,1 dfw.net,48 |
| | dhp.com,1 |
| a | dmi.net,12 |
| - | etca.fr,1 |
| | frl.com,2 fsd.com,45 |
| | ansa.ch,12 |
| | gs1.con,35 |
| | gtx.com,35 |
| ۶_ | hal.COM,73 ids.net,36 |
| | inec.be, 151 |
| | j51.com,1 |
| | jiop.fi,4 |
| - | k2t.com,11 |
| | lca.com,4 lpa.com,7 |
| | nak.con,1 |
| | mne.net,12 |
| | nfe.com,42 |
| | nova.ca,66 ols.com,2 |
| | ols.com,1 |
| | pie.org,12 |
| - | rdr.com,10 |
| Control | rto.com,1 scn.oro.36 |
| | 361.010.20 |

| Ubuntu (02122017) [Running] - Oracle VM VirtualBox part-r-00000 (-/Downloads) - gedit |
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| part-r-00000 (-/Downloads) - gedic |
| Open - FI |
| ids.net,36 |
| [mec.be,151 j51.com,1 |
| jiop.fi,4 |
| k2t.com,11 lca.com,4 |
| lpa.con,7 |
| nak.con,1 |
| nne.net,12 nfe.com,42 |
| nova.ca,66 |
| ois.com,2 ols.com,1 |
| fpie.org,12 |
| rdr.com,10 rio.com,1 |
| scn.org,36 |
| 🖽 sdt.con,344 simi.is,64 |
| tds.com,247 |
| tfs.com,26 twp.com,4 |
| upx.net,13 |
| utb.edu,1 |
| vrb.com,2 zai.com,1 |
| 2 zen.con,36 |
| 4CTY.ORG, 3 MAIN.COM, 3 |
| PEAK.ORG, 32 |
| Sony.COM, 59 XULA.EDU, 2 |
| andi.com,34 |
| apci.com,161 apl.eecs,1 |
| apti.com,45 |
| ares.com,12 |
| e asae.org,5 bgis.com,44 |
| cnct.com,321 |
| ecnet.ec,7 elvis.ru,1 |
| fi.hh.se,92 |
| gcon.com,2 hl.db.dk,10 |
| ined.com,35 |
| inch.com,15 iota.com,18 |
| jart.com,30 |
| kelk.com,12 lafn.org,24 |
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| Ubuntu (02122017) [Running] - Oracle VM VirtualBox |
|--|
| part-r-00000 (-/Downloads) - gedit |
| Open - A |
| kelk.con,12 |
| lafn.org,24 mktg.com,1 |
| moli.com,17 |
| nuhos.fi,1 netc.com,5 |
| netti.fi,23 |
| oxhp.com,2 |
| peri.org,10 peri.com,19 |
| pi.hh.se,7 |
| ppco.com,29 pwcm.com,25 |
| 🖌 raba.con,17 |
| rotol.fi,24 sbcm.com,94 |
| sdrc.com,267 |
| sjfc.edu,1 tals.com,68 |
| test.com,24 |
| tgis.com,52 tl.dr.dk,29 |
| uwasa.fi,20 |
| veng.com,56 vssp.com,2 |
| well.com,12 |
| a yebb.com,7 127.0.0.1,3 |
| 144.3.2.2,7 |
| 152.2.0.3,19 196.3.0.2,3 |
| 196.3.3.5,25 |
| 196.4.0.3,28 206.8.2.8,1 |
| 2- 35.dx.com,28 |
| 36.42.0.8,65 38.12.1.6,9 |
| 400gtw.nl,83 |
| 41.dx.com,1 6.21.13.8,31 |
| 6.36.3.36,32 |
| 6.37.7.36,11 ,1 |
| AC.Dal.Ca,11 |
| CMPC0.COM,29 GS.ulb.no,35 |
| IS.Dal.Ca,35 |
| ISTINC.CA,7 |
| TRENTU.CA,16 The-B.org,1 |
| aetna.com,1 |
| |

| 💐 Ubuntu (02122017) [Running] - Oracle VM VirtualBox |
|--|
| part-r-00000 (~/Downloads) - gedit |
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| Open - II |
| clark.edu,2 |
| clark.net,5607 |
| clinet.fi,1 |
| cpcug.org,232 cs.bc.edu,1 |
| dames.com,3 |
| db.sw.com,1 |
| dmgow.com,8 |
| ee.hit.fi,1 |
| ee.tut.fi,2 eldec.com,42 |
| ettec.com,42 emtek.com,6 |
| fangz.com,1 |
| fasor.com,2 |
| fw.itu.ch,1 |
| g8.rmc.ca,34 |
| genie.com,77 glock.com,11 |
| gsv.gu.se,234 |
| hq.si.net,13 |
| huber.com,130 |
| idiaiac.dk,99 |
| iainc.com,51 iglou.com,53 |
| inmax.com,21 |
| innet.com,2 |
| isgate.is,224 |
| itp.pp.fi,9 |
| lacma.org,10 largo.com,2 |
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