

دیدنش با چشم چون ممکن نبود - اندر آن تاریکیش کف میبسود

مولانا



INTRODUCTION TO HADOOP

Big Data, Hadoop & more Milad As (Ravexina)

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SAVE BUTTON?



FLOPPY DISK



BUT THINGS CHANGED!







INCREASE IN

Sources (Users, Clients), Forms



FLOOD IS COMING





MySQL, MariaDB, Access, SQL Server, Oracle Time out on expensive servers...

OVERLOAD

You must change your technologies



WHAT MAKES BIG DATA



Implicit and Explicit



Forums, Messengers, Comments, Q/A, Reviews



Share ...!? Did you just pause that?



Share, view, follow



THESE ARE ONLY THE VISIBLE PARTS OF BIGDATA



Applications

Applications

Servers (Web, Email, Proxy, ...)

Applications Servers (Web, Email, Proxy, ...) Systems

Applications

Servers (Web, Email, Proxy, ...)

Systems journalct

Hardwares Access point

Applications Servers (Web, Email, Proxy, ...) Systems _{journalct} Hardwares _{Access point} Life

Applications

Servers (Web, Email, Proxy, ...)

Systems journalct

Hardwares Access point

Clicks

Applications

Servers (Web, Email, Proxy, ...)

Systems journalct

Hardwares Access point

Clicks

Applications

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Systems

Hardwares Access point

Clicks

Applications

Servers (Web, Email, Proxy, ...)

Systems

Hardwares Access point

MAIN SOURCE

Mysterious and valuable part!



DIGITAL FOOTPRINT



A simple tweet

HOW CAN WE PROCESS THESE DATA?

SCALABILITY





Low resistance but inexpensive and infinite


Organizations

SATURATION



Bottlenecks like: Disk IO, load avg

No matter how much memory I have...

SATURATION



Bottlenecks like: Disk IO, load avg

No matter how much memory I have...



Now we know what we are looking for!

CLIENT SERVER ARCHITECTURE



RAID



GOOGLE



GFS, ???

DOUG CUTTING



Apache Lucene (IR) Apache Nutch (Web Crawler)

WHITE PAPER



MAPREDUCE



It's just a framework, C4.5



HADOOP COMES INTO PLAY!



NAMING



HORTONWORKS DEFINATION

An opensource software platform for distributed storage and distributed processing of very large data sets on computer clusters built from commodity hardware

TWO SIDES OF HADOOP



TWO SIDES OF HADOOP



GFS, MapReduce

TWO SIDES OF HADOOP



GFS, MapReduce HDFS, Hadoop MapReduce

WHAT WE DO WITH DATA

WHAT WE DO WITH DATA

- Save
- Transfer
- Join
- Index
- Analytics
- Aggregate
- Visualize

ECOSYSTEM



ECOSYSTEM



HDFS

Allows us to distribute the storage All hard drives look likes a single huge hard disk Keeps copy of data No single point of failure

YARN

Manage the resouces What get to run tasks and when Which node is available We build applications on top of it

MAPREDUCE

Got remvoed from YARN

PIG

No java, Python? Scripting language like SQL Transforms the script to something than can be run on MapReduce

HIVE

SQL Makes the data to look like a RDBMS

TEZ

Hive on TEZ is faster than MapReduce

SPARK

Sitting at same level of MapReduce on top of YARN Or **MESOS** Python, Java, Scala Fast Active development Handle SQL Query Machine learning Handle Stream data

STORM

Processing streaming data Sensors, Logs Spark streaming does the same thing Update machine learning model Update data as it comes

OOZIE SCHEDULE OF JOBS

Complicated steps Load to hive, Query using spark then transform to HBASE.

ZOOKEEPER

Cordinates everything on clusters Which node is up or down Many of these apps relay on zookeeper

DATA INGESTION SQOOP

Turn into hadoop, talks to ODBS JDBC FLUME

Transport Web logs into hadoop (spark, storm) KAFKA

Like flume but more general, cluster of PC or webservers or whatever to broadcast into Hadoop cluster.

DISTRIBUTIONS



HORTONWORKS SANDBOX



AN HIVE EXAMPLE
TTY

HDP 2.5 http://hortonworks.com

To initiate your Hortonworks Sandbox session, please open a browser and enter this address in the browser's address field: http://127.0.0.1:8888/

Log in to this virtual machine: Linux/Windows <Alt+F5>, Mac OS X <Fn+Alt+F5>

APACHE AMBARI

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O HDFS	Metrics • Heatmaps				
YARN					
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🗢 HBase	31%	1998/2000	2000 DataNodes		
Hive			More *		
WebHCat					L 1
Falcon	CPU Usage	Cluster Load	NameNode Heap	NameNode RPC	NameNode CPU WIO
Storm	100%				
Oczie			21%	0.06 ms	0.0%
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movie_id	rating	rating_time
INT -	INT -	
242	3	881250949
302	3	891717742
377	1	878887116
51	2	880606923
346	1	886397596
474	4	884182806

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	(GoldenEye	(1995)			01-Jar	n-1995				

Four Rooms (1995) 01-Jan-1995

	•	Worksheet	
		<pre>1 SELECT movie_id, count(movie_id) as ratingCount</pre>	
bles		2 FROM ratings 3 GROUP BY movie_id	
ses		4 ORDER BY ratingCount 5 DESC;	
names			
id	INT		
	STRING		
n3	STRING		
n4	STRING		
n5	STRING		
n6	INT		
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```
movie_id, count(movie_id) as ratingCount
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DATA VISUALIZER

Data Visualization	Data Exp	plorer				Maximum Row Count: 10000	OK
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STAR WARS

Worksheet	
1 SELECT name 2 FROM movie_names 3 WHERE movie_id = 50;	
Execute Explain Save as	New Worksheet
\searrow	
Query Process Results (Status: SUCCEEDED)	Save results
Logs Results	
Filter columns	previous next
name	
Star Wars (1977)	







Hadoop V2 separated from MapReduce



Hadoop V2 separated from MapReduce Run MapReduce alternatives on it (TEZ)



Hadoop V2 separated from MapReduce Run MapReduce alternatives on it (TEZ) Idea? split the computation across the cluster



Hadoop V2 separated from MapReduce Run MapReduce alternatives on it (TEZ) Idea? split the computation across the cluster Maintain data locality (Integrated with HDFS, where data lives)

YARN



YARN



YARN



We can have multiple resource manager

TEZ (STILL YARN)







Allow our big data to be stored across entire cluster in distributed and reliable manner.



Allow our big data to be stored across entire cluster in distributed and reliable manner.

Handling large files



Allow our big data to be stored across entire cluster in distributed and reliable manner. Handling large files Breaking data into blocks - 128 MB



Allow our big data to be stored across entire cluster in distributed and reliable manner. Handling large files Breaking data into blocks - 128 MB Keeps multiple copies of these blocks (Clever way)



Allow our big data to be stored across entire cluster in distributed and reliable manner. Handling large files Breaking data into blocks - 128 MB Keeps multiple copies of these blocks (Clever way) Allow us to use regular computers (No special hardware needed)

HDFS



HA HDFS



Client <---- > Zookeeper One namenode is active at a time

HDFS FEDERATION



Sub Directories -> namespace Volume -> each namenode manage one namespace volume

HDFS FEDERATION



Sub Directories -> namespace Volume -> each namenode manage one namespace volume







Map: Transfer Data that we care about



Map: Transfer Data that we care about

Reduce: Aggregate Data



Map: Transfer Data that we care about Shuffle and sort Reduce: Aggregate Data








def reducer_count_ratings(self, key, values):
 yield key, sum(values)

```
from mrjob.job import MRJob
from mrjob.step import MRStep
class RatingsBreakdown(MRJob):
    def steps(self):
        return [
            MRStep(mapper=self.mapper_get_ratings,
                   reducer=self.reducer_count_ratings)
        ]
    def mapper_get_ratings(self, _, line):
        (userID, movieID, rating, timestamp) = line.split('\t')
        yield rating, 1
    def reducer_count_ratings(self, key, values):
        yield key, sum(values)
if __name__ == '__main__':
   RatingsBreakdown.run()
```

