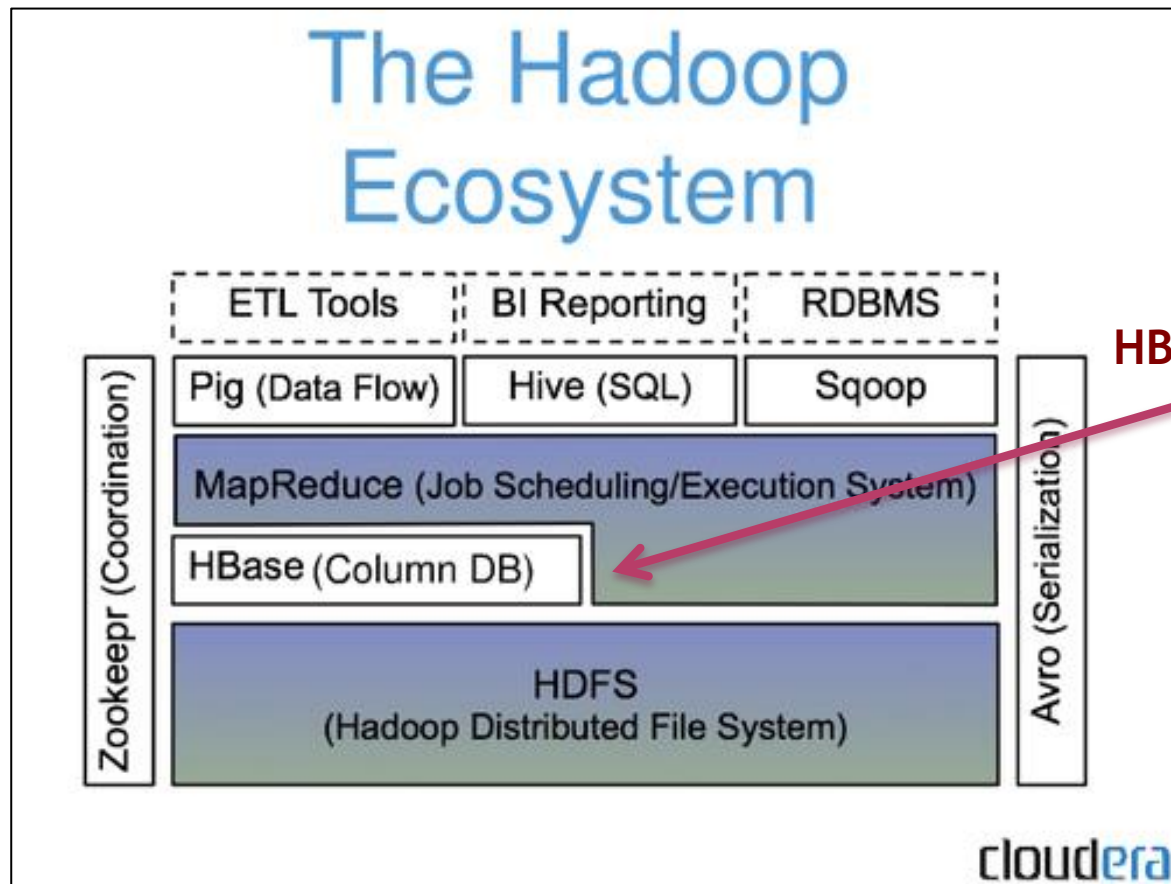


# CASE STUDY: HADOOP

# OUTLINE

- ◉ Hadoop - Basics
- ◉ HDFS
  - Goals
  - Architecture
  - Other functions
- ◉ MapReduce
  - Basics
  - Word Count Example
  - Handy tools
  - Finding shortest path example
- ◉ Related Apache sub-projects (Pig, HBase, Hive)

# HBASE: PART OF HADOOP'S ECOSYSTEM



HBase is built on top of HDFS



HBase files are internally stored in HDFS

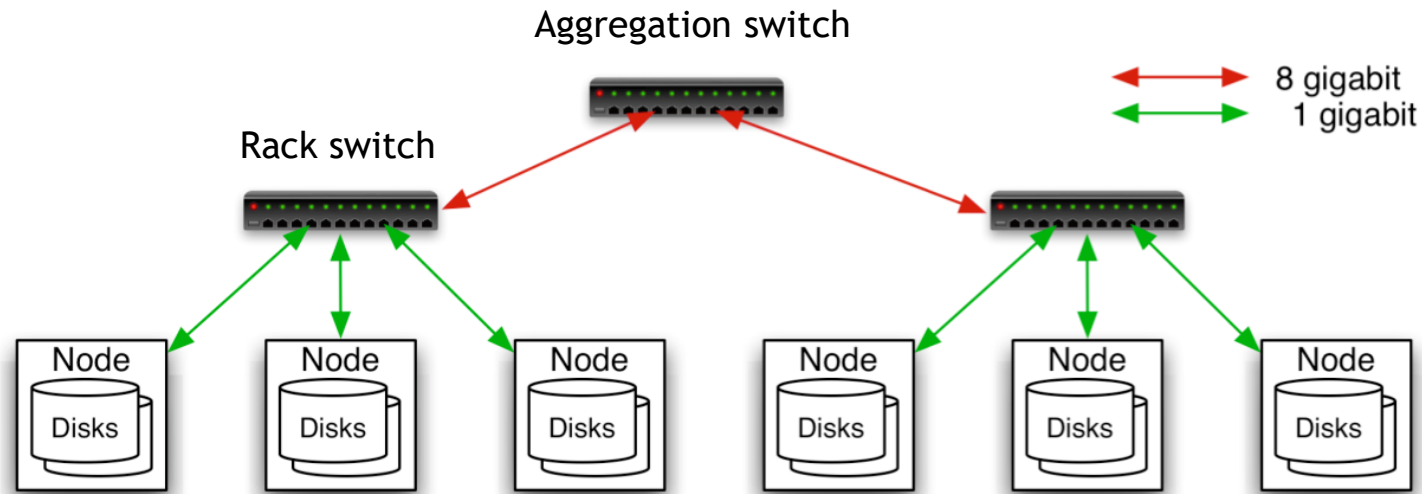
# HADOOP - WHY ?

- Need to process huge datasets on large clusters of computers
- Very expensive to build reliability into each application
- Nodes fail every day
  - Failure is expected, rather than exceptional
  - The number of nodes in a cluster is not constant
- Need a common infrastructure
  - Efficient, reliable, easy to use
  - Open Source, Apache Licence

# WHO USES HADOOP?

- ◉ Amazon/A9
- ◉ Facebook
- ◉ Google
- ◉ New York Times
- ◉ Veoh
- ◉ Yahoo!
- ◉ .... many more

# COMMODITY HARDWARE



## Typically in 2 level architecture

- Nodes are commodity PCs
- 30-40 nodes/rack
- Uplink from rack is 3-4 gigabit
- Rack-internal is 1 gigabit

# HADOOP DISTRIBUTED FILE SYSTEM (HDFS)

# GOALS OF HDFS

- Very Large Distributed File System
  - 10K nodes, 100 million files, 10PB
- Assumes Commodity Hardware
  - Files are replicated to handle hardware failure
  - Detect failures and recover from them
- Optimized for Batch Processing
  - Data locations exposed so that computations can move to where data resides
  - Provides very high aggregate bandwidth



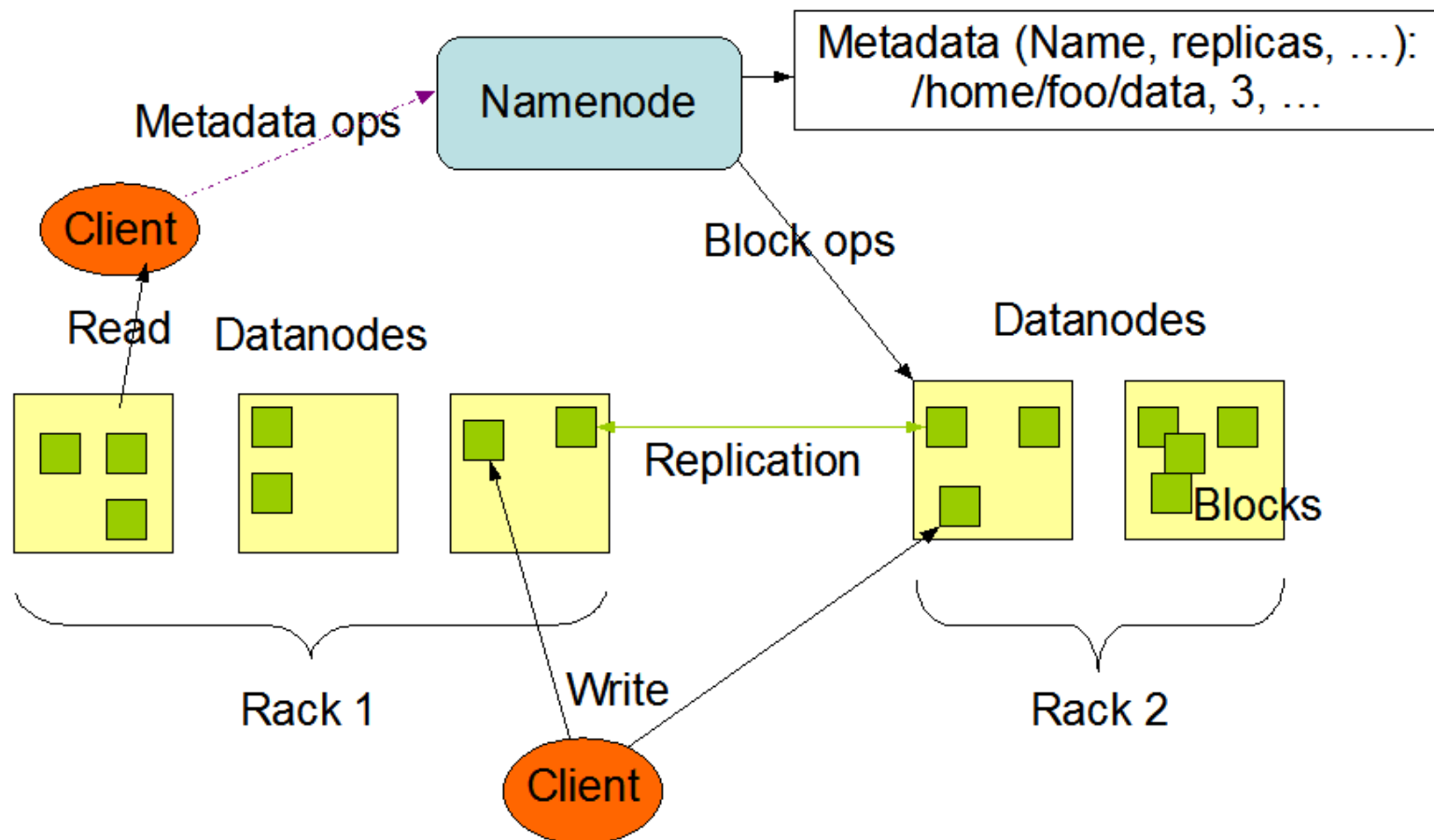


# DISTRIBUTED FILE SYSTEM

- ◉ Single Namespace for entire cluster
- ◉ Data Coherency
  - Write-once-read-many access model
  - Client can only append to existing files
- ◉ Files are broken up into blocks
  - Typically 64MB block size
  - Each block replicated on multiple DataNodes
- ◉ Intelligent Client
  - Client can find location of blocks
  - Client accesses data directly from DataNode

# HDFS ARCHITECTURE

## HDFS Architecture



# FUNCTIONS OF A NAMENODE

- ◉ **Manages File System Namespace**
  - Maps a file name to a set of blocks
  - Maps a block to the DataNodes where it resides
- ◉ **Cluster Configuration Management**
- ◉ **Replication Engine for Blocks**

# NAMENODE METADATA

- **Metadata in Memory**

- The entire metadata is in main memory
- No demand paging of metadata

- **Types of metadata**

- List of files
- List of Blocks for each file
- List of DataNodes for each block
- File attributes, e.g. creation time, replication factor

- **A Transaction Log**

- Records file creations, file deletions etc

# DATANODE

## ⦿ A Block Server

- Stores data in the local file system (e.g. ext3)
- Stores metadata of a block (e.g. CRC)
- Serves data and metadata to Clients

## ⦿ Block Report

- Periodically sends a report of all existing blocks to the NameNode

## ⦿ Facilitates Pipelining of Data

- Forwards data to other specified DataNodes

# BLOCK PLACEMENT

- Current Strategy
  - One replica on local node
  - Second replica on a remote rack
  - Third replica on same remote rack
  - Additional replicas are randomly placed
- Clients read from nearest replicas
- Would like to make this policy pluggable

# HEARTBEATS

- DataNodes send heartbeat to the NameNode
  - Once every 3 seconds
- NameNode uses heartbeats to detect DataNode failure

# REPLICATION ENGINE

- NameNode detects DataNode failures
  - Chooses new DataNodes for new replicas
  - Balances disk usage
  - Balances communication traffic to DataNodes



# DATA CORRECTNESS

- ⦿ Use Checksums to validate data
  - Use CRC32
- ⦿ File Creation
  - Client computes checksum per 512 bytes
  - DataNode stores the checksum
- ⦿ File access
  - Client retrieves the data and checksum from DataNode
  - If Validation fails, Client tries other replicas

# NAMENODE FAILURE

- ◉ A single point of failure
- ◉ Transaction Log stored in multiple directories
  - A directory on the local file system
  - A directory on a remote file system (NFS/CIFS)
- ◉ Need to develop a real HA solution

# SECONDARY NAMENODE

- Copies FsImage and Transaction Log from Namenode to a temporary directory
- Merges FSImage and Transaction Log into a new FSImage in temporary directory
- Uploads new FSImage to the NameNode
  - Transaction Log on NameNode is purged

# USER INTERFACE

## ⦿ Commads for HDFS User:

- `hadoop dfs -mkdir /foodir`
- `hadoop dfs -cat /foodir/myfile.txt`
- `hadoop dfs -rm /foodir/myfile.txt`

## ⦿ Commands for HDFS Administrator

- `hadoop dfsadmin -report`
- `hadoop dfsadmin -decommision datanodename`

## ⦿ Web Interface

- `http://host:port/dfshealth.jsp`

PIG

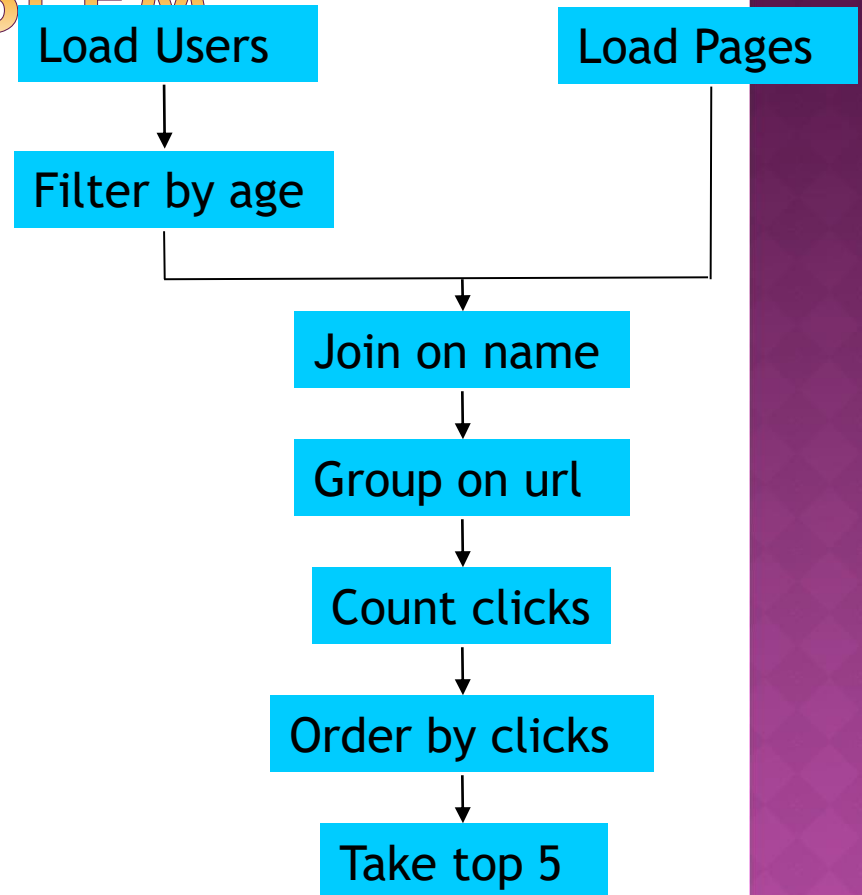
# PIG

- ◉ Started at Yahoo! Research
- ◉ Now runs about 30% of Yahoo!'s jobs
- ◉ Features
  - Expresses sequences of MapReduce jobs
  - Data model: nested “bags” of items
  - Provides relational (SQL) operators (JOIN, GROUP BY, etc.)
  - Easy to plug in Java functions



# AN EXAMPLE PROBLEM

- Suppose you have user data in a file, website data in another, and you need to find the top 5 most visited pages by users aged 18-25

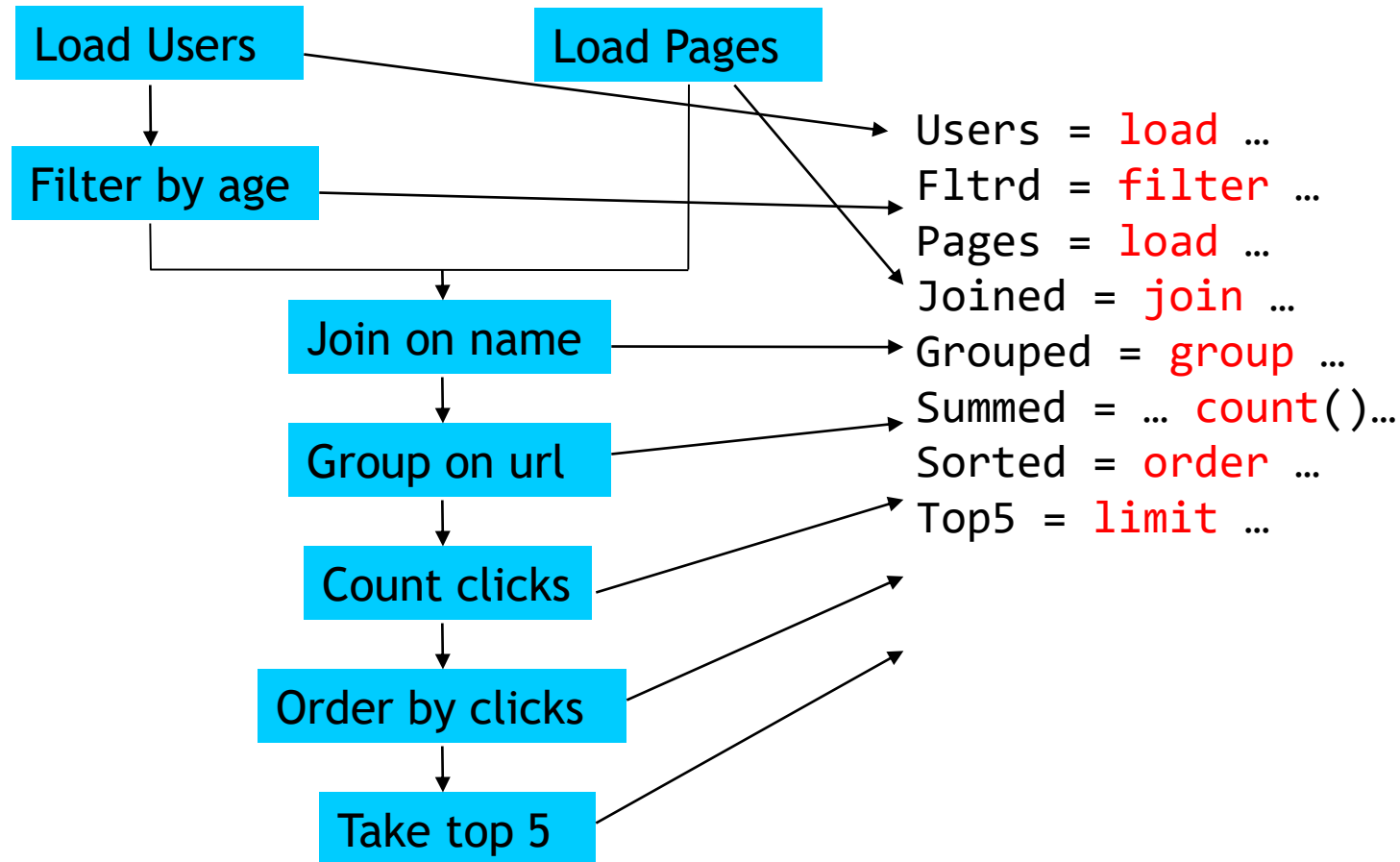


# IN PIG LATIN

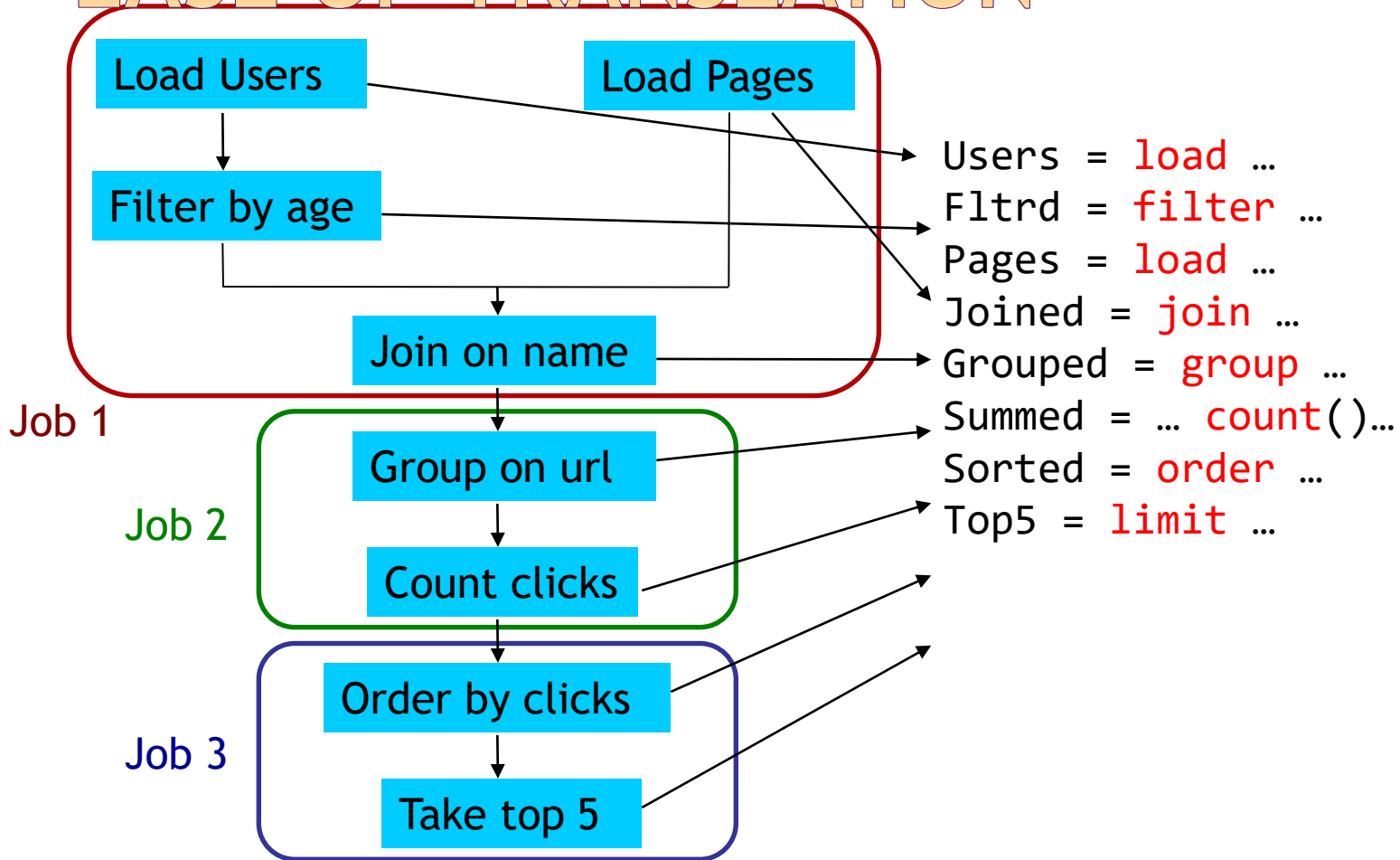
```
Users = load 'users' as (name, age);
Filtered = filter Users by age >= 18 and age <=
    25;
Pages = load 'pages' as (user, url);
Joined = join Filtered by name, Pages by user;
Grouped = group Joined by url;
Summed = foreach Grouped generate group,
    count(Joined) as clicks;
Sorted = order Summed by clicks desc;
Top5 = limit Sorted 5;
store Top5 into 'top5sites';
```



# EASE OF TRANSLATION



# EASE OF TRANSLATION



**HBASE**

# HBASE - WHAT?

- ◉ Modeled on Google's Bigtable
- ◉ Row/column store
- ◉ Billions of rows/millions on columns
- ◉ Column-oriented - nulls are free
- ◉ Untyped - stores byte[]

# HBASE - DATA MODEL

Row	Timestamp	Column family: animal:		Column family repairs:
		animal:type	animal:size	repairs:cost
enclosure1	t2	zebra		1000 EUR
	t1	lion	big	
enclosure2	...	...	...	...

# HBASE - DATA STORAGE

Column family animal:

(enclosure1, t2, animal:type)	zebra
(enclosure1, t1, animal:size)	big
(enclosure1, t1, animal:type)	lion

Column family repairs:

(enclosure1, t1, repairs:cost)	1000 EUR
--------------------------------	----------

# HBASE - CODE

*HTable table = ...*

*Text row = new Text("enclosure1");*

*Text col1 = new Text("animal:type");*

*Text col2 = new Text("animal:size");*

*BatchUpdate update = new BatchUpdate(row);*

*update.put(col1, "lion".getBytes("UTF-8"));*

*update.put(col2, "big".getBytes("UTF-8"));*

*table.commit(update);*

*update = new BatchUpdate(row);*

*update.put(col1, "zebra".getBytes("UTF-8"));*

*table.commit(update);*

# HBASE - QUERYING

- ◉ Retrieve a cell

Cell =

```
table.getRow("enclosure1").getColumn("animal:type").getValue();
```

- ◉ Retrieve a row

```
RowResult = table.getRow( "enclosure1" );
```

- ◉ Scan through a range of rows

```
Scanner s = table.getScanner( new String[] { "animal:type" } );
```



HIVE

# HIVE

- ◉ Developed at Facebook
- ◉ Used for majority of Facebook jobs
- ◉ “Relational database” built on Hadoop
  - Maintains list of table schemas
  - SQL-like query language (HiveQL)
  - Can call Hadoop Streaming scripts from HiveQL
  - Supports table partitioning, clustering, complex data types, some optimizations

The Hive logo, featuring the word "Hive" in a white, sans-serif font on a dark blue rectangular background.

# CREATING A HIVE TABLE

```
CREATE TABLE page_views(viewTime INT, userid BIGINT,  
                           page_url STRING, referrer_url STRING,  
                           ip STRING COMMENT 'User IP address')  
COMMENT 'This is the page view table'  
PARTITIONED BY(dt STRING, country STRING)  
STORED AS SEQUENCEFILE;
```

- Partitioning breaks table into separate files for each (dt, country) pair

Ex: /hive/page\_view/dt=2008-06-08,country=USA  
/hive/page\_view/dt=2008-06-08,country=CA

## A SIMPLE QUERY

- Find all page views coming from xyz.com on March 31<sup>st</sup>:

```
SELECT page_views.*  
FROM page_views  
WHERE page_views.date >= '2008-03-01'  
AND page_views.date <= '2008-03-31'  
AND page_views.referrer_url like '%xyz.com';
```

- Hive only reads partition 2008-03-01, \* instead of scanning entire table

# AGGREGATION AND JOINS

- Count users who visited each page by gender:

```
SELECT pv.page_url, u.gender, COUNT(DISTINCT u.id)
FROM page_views pv JOIN user u ON (pv.userid = u.id)
GROUP BY pv.page_url, u.gender
WHERE pv.date = '2008-03-03';
```

- Sample output:

page_url	gender	count(userid)
home.php	MALE	12,141,412
home.php	FEMALE	15,431,579
photo.php	MALE	23,941,451
photo.php	FEMALE	21,231,314

# USING A HADOOP STREAMING MAPPER SCRIPT

```
SELECT TRANSFORM(page_views.userid,  
                  page_views.date)  
USING 'map_script.py'  
AS dt, uid CLUSTER BY dt  
FROM page_views;
```

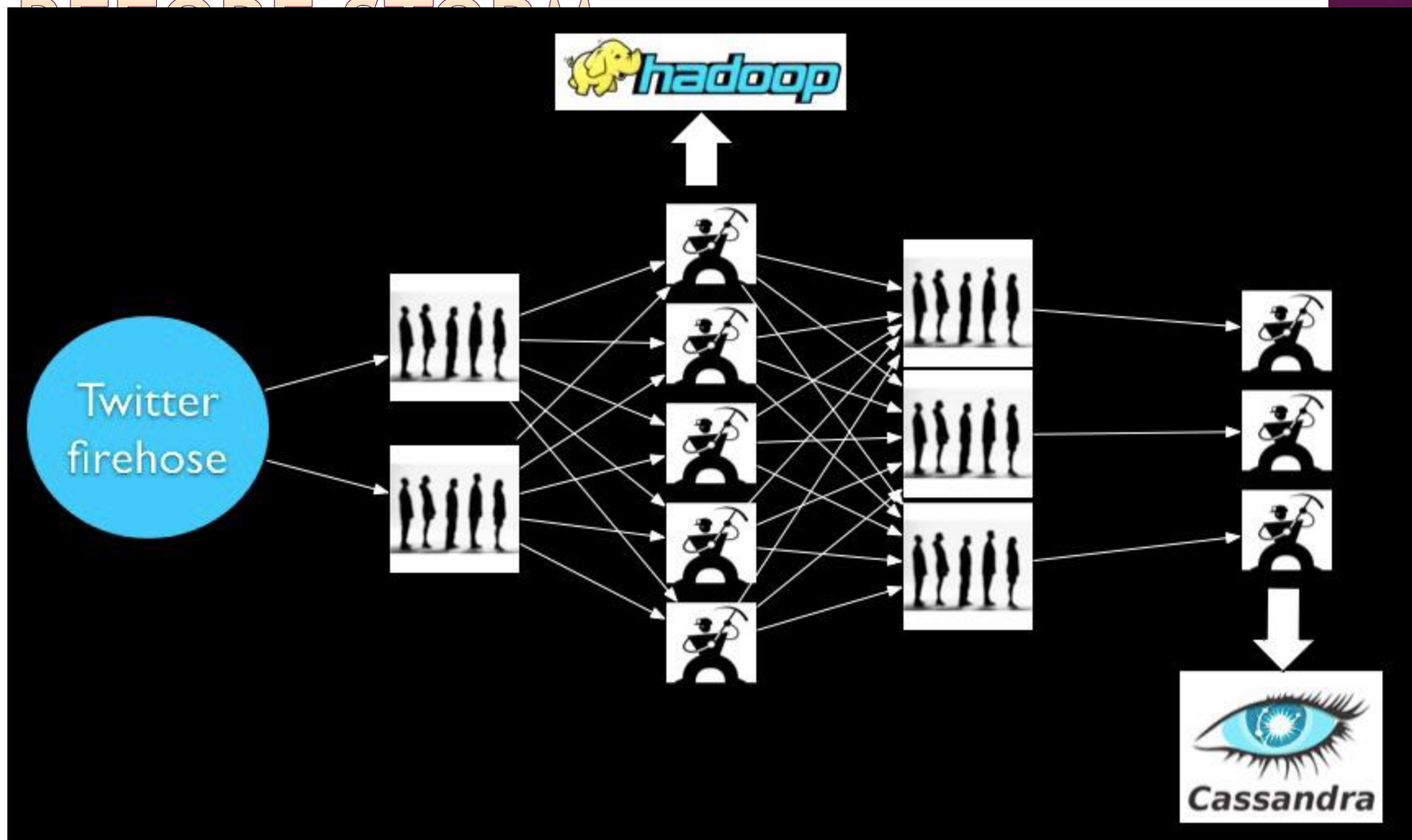
**STORM**

# STORM

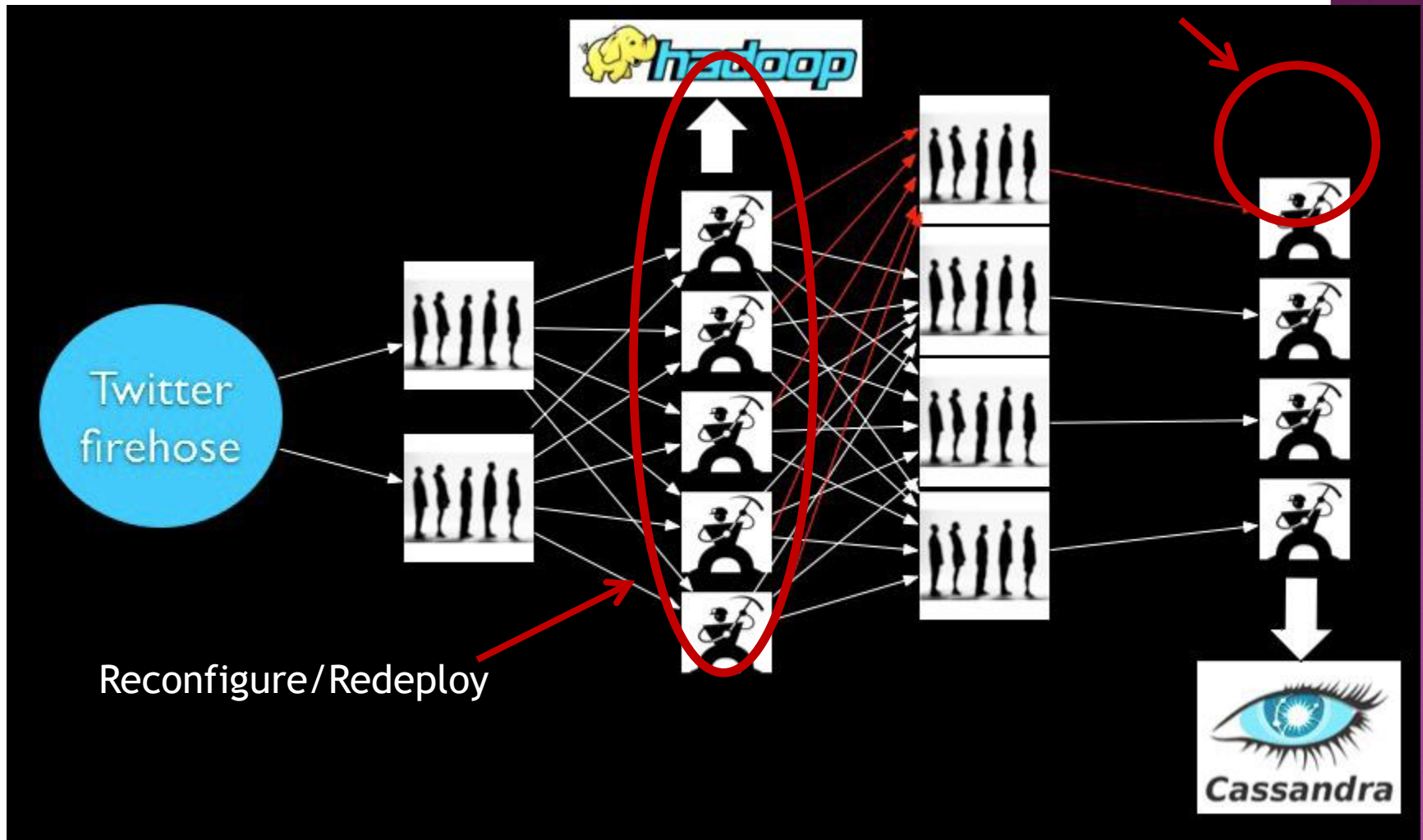
- ◉ Developed by BackType which was acquired by Twitter
- ◉ Lots of tools for data (i.e. batch) processing
  - Hadoop, Pig, HBase, Hive, ...
- ◉ None of them are realtime systems which is becoming a real requirement for businesses
- ◉ Storm provides realtime computation
  - Scalable
  - Guarantees no data loss
  - Extremely robust and fault-tolerant
  - Programming language agnostic







# BEFORE STORM - ADDING A WORKER



# PROBLEMS

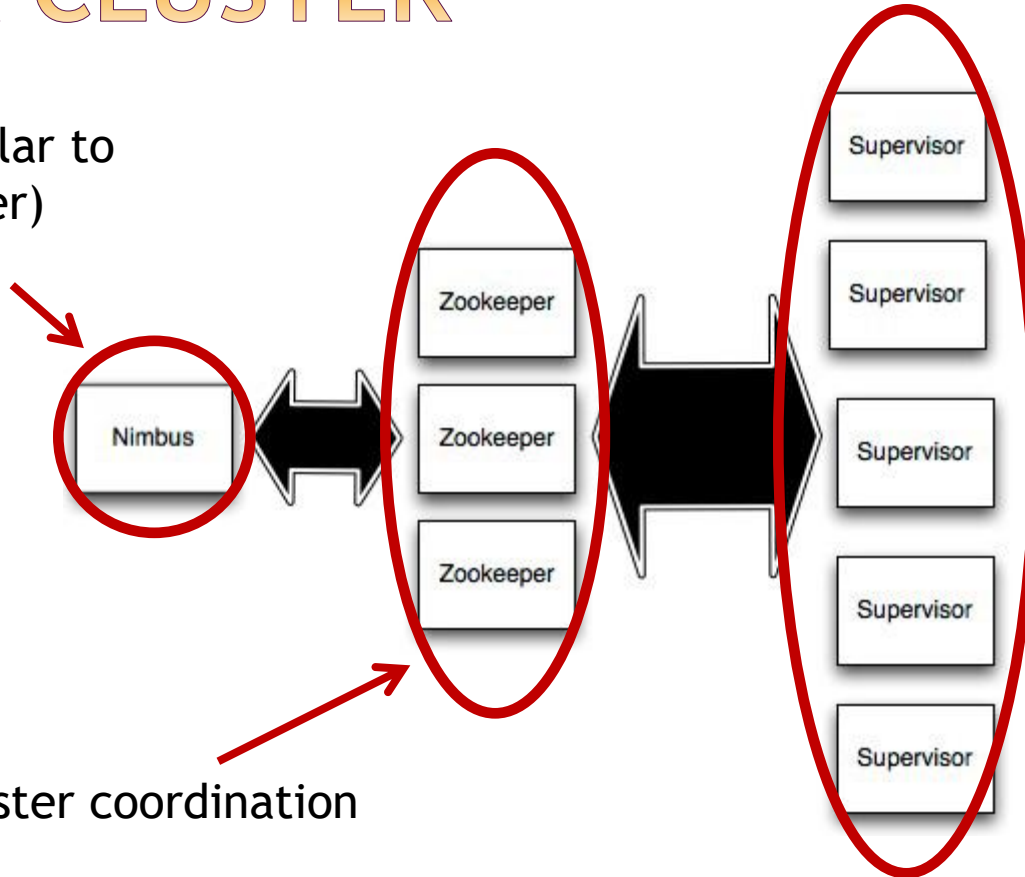
- ◉ Scaling is painful
- ◉ Poor fault-tolerance
- ◉ Coding is tedious

# WHAT WE WANT

- ◉ Guaranteed data processing
- ◉ Horizontal scalability
- ◉ Fault-tolerance
- ◉ No intermediate message brokers!
- ◉ Higher level abstraction than message passing
- ◉ “Just works” !!

# STORM CLUSTER

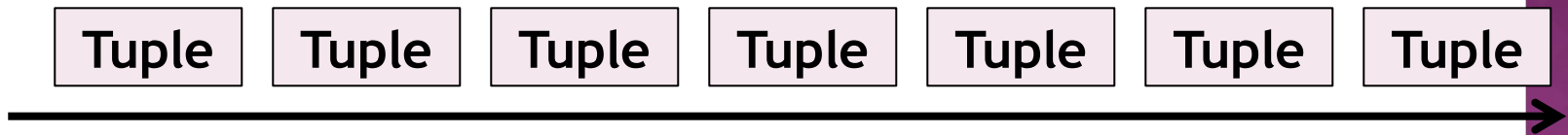
Master node (similar to Hadoop JobTracker)



Used for cluster coordination

Run worker processes

# STREAMS



Unbounded sequence of tuples