

SQL on Hadoop - Analyzing Big Data with Hive

Ahmad Alkilani
www.pluralsight.com



pluralsight
hardcore developer training

Introduction to Hadoop

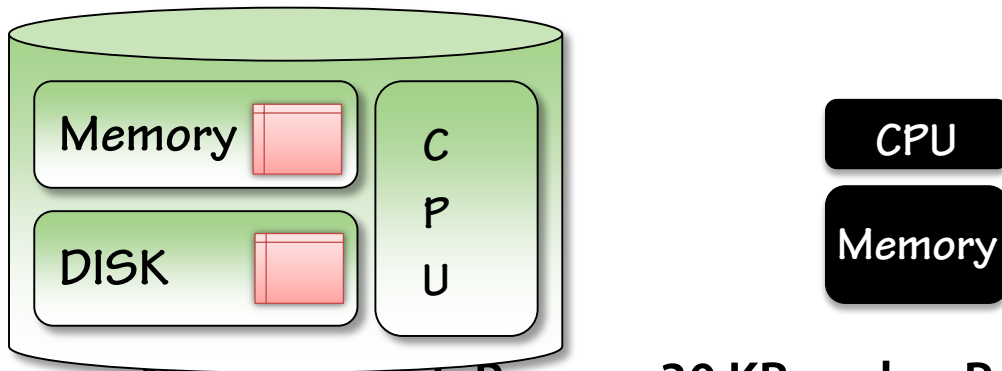
Ahmad Alkilani
www.pluralsight.com



Outline

- **Why Hadoop? Motivation**
- **Hadoop architecture and distributed computing**
- **HDFS**
- **MapReduce**
- **Getting up and running**

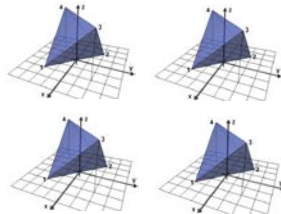
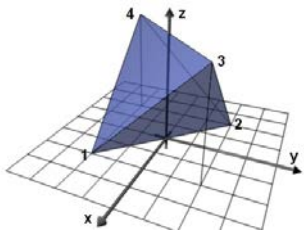
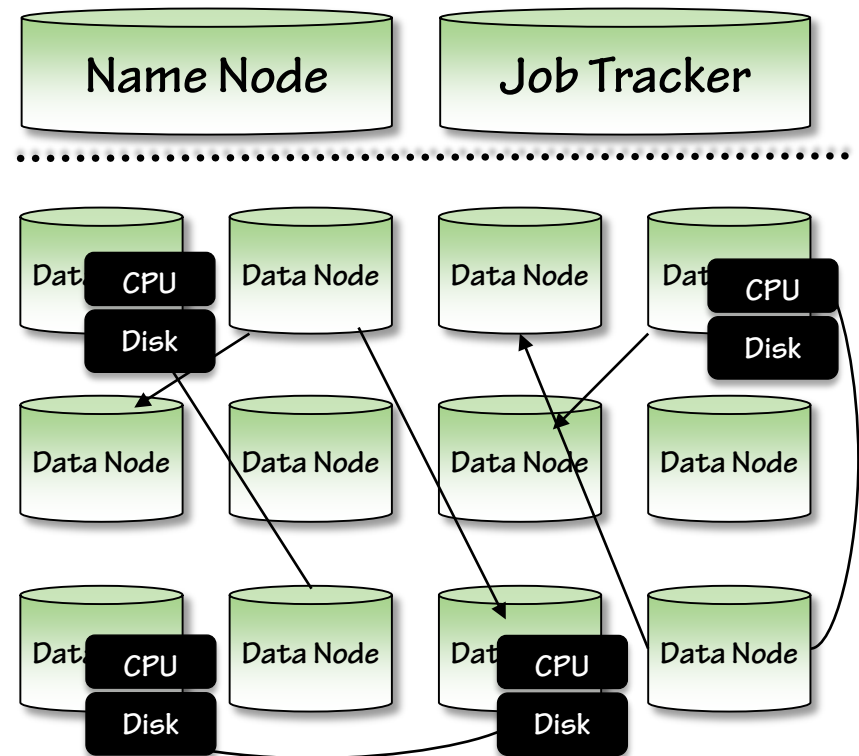
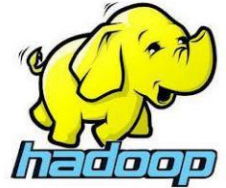
Motivation for Hadoop



- ~40 Billion Web Pages x 30 KB each = Petabyte
- Today's average disk speed reads about 120 MB/sec
- Little over 3 months to read the web!
- Approximately 1,000 drives to store and use

Distributed Computing Challenges

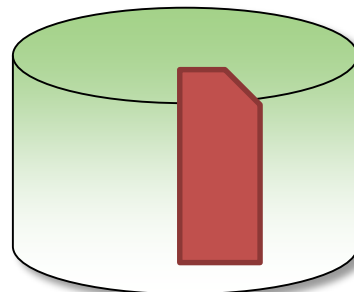
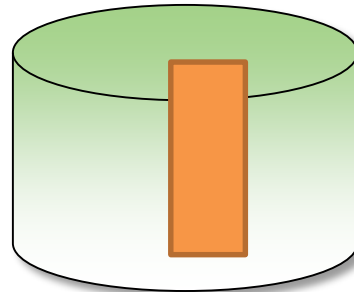
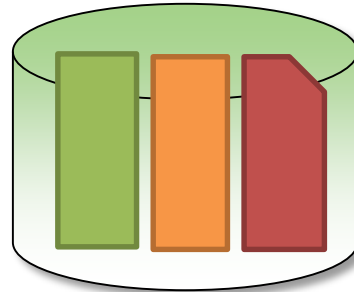
- Scale out with distributed computing
- Hadoop based on Google's implementation
- Volume, Velocity, and Variety
- Recover from failures
- Shared nothing architecture
- Hadoop file system (HDFS)
- MapReduce



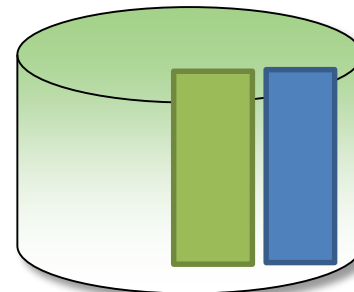
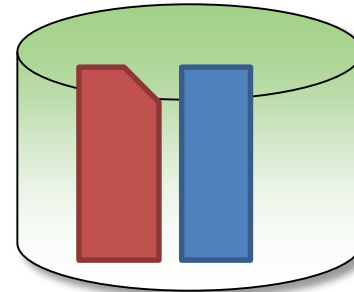
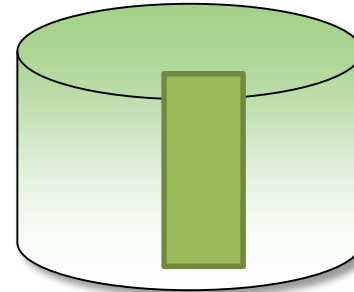
Hadoop File System (HDFS)



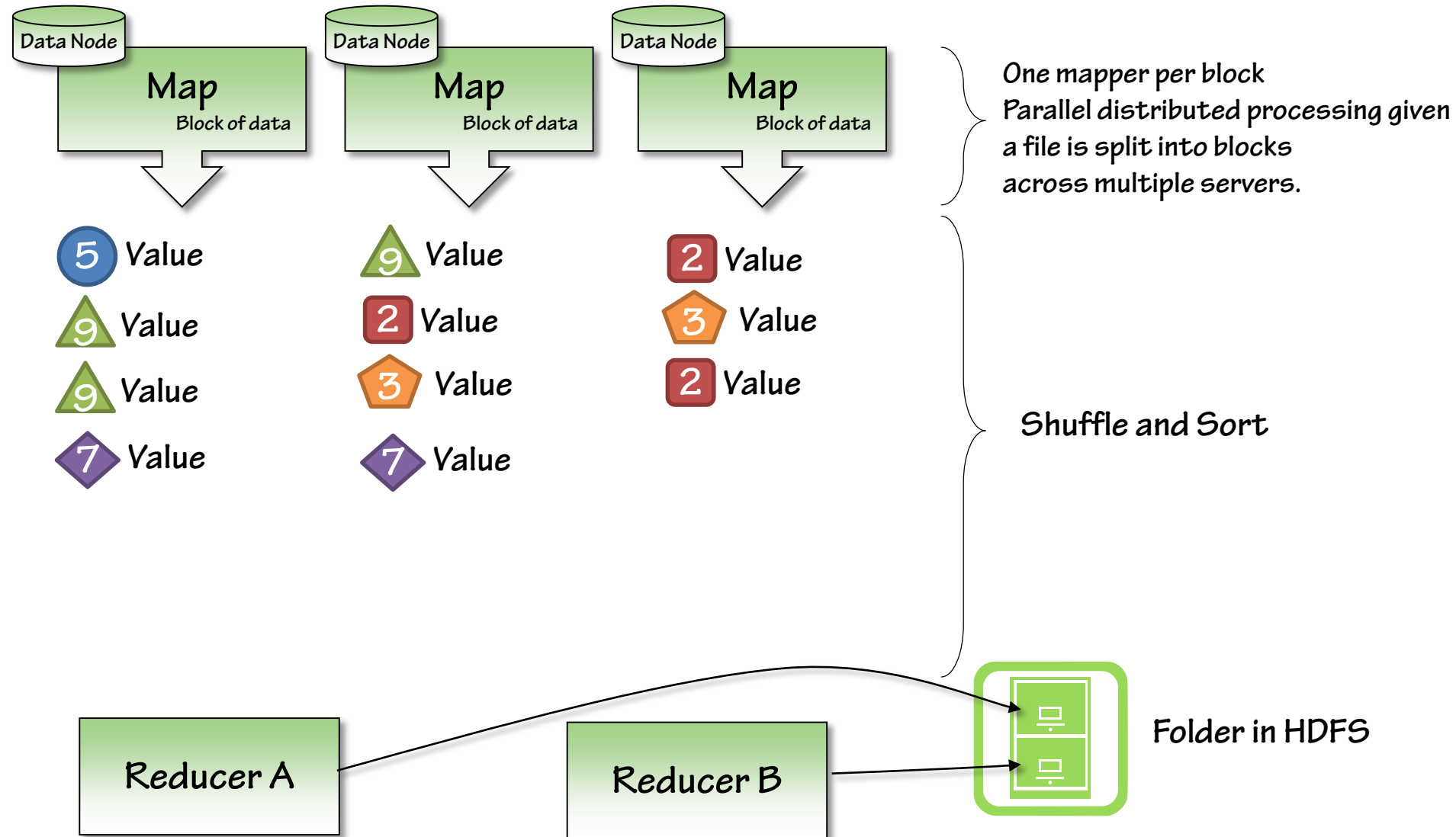
Server Rack A



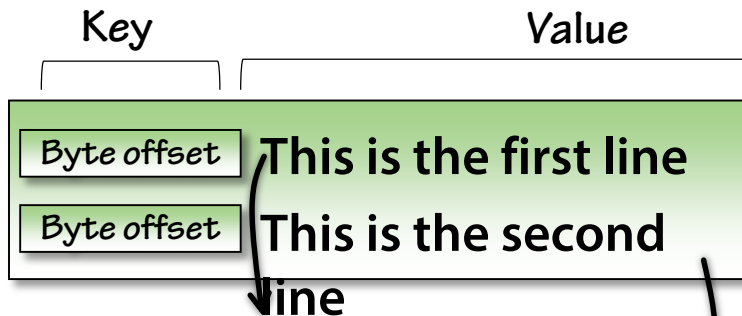
Server Rack B



MapReduce



Word Count Example



Key	Value
This	1
is	1
the	1
first	1
line	1

Key	Value
This	1
is	1
the	1
second	1
line	1

```
String line = value.toString();
StringTokenizer tokenizer = new StringTokenizer(line);
while (tokenizer.hasMoreTokens())
{
    word.set(tokenizer.nextToken());
    context.write(word, one);
}
```

Key	Value
This	1
This	1
the	1
the	1
second	1
first	1

Key	Value
line	1
line	1
is	1
is	1

Reducer A

first 1
second 1
the 2
This 2

Reducer B

is 2
line 2

```
int sum = 0;
for (IntWritable val : values)
{ sum += val.get(); }
context.write(key, new IntWritable(sum));
```


Basic commands using HDFS

Hadoop Demo

Environment Setup

- **Course focus is on development**
- **Use a Virtual Machine image to follow along with examples**
- **Pseudo distributed sandbox**
 - Replication factor set to 1
 - Name Node, Job Tracker, Data Node, and Task Tracker on a single machine
- **Demos using Hortonworks' HDP sandbox**
 - Hive 0.10, 0.11 and above

Summary

- **Distributed computing and scaling out to solve big data problems**
- **Key system characteristics**
 - Built to handle failures
 - Move processing to the data
 - Failures are inevitable. Embracing this allows for solutions built on commodity servers
- **MapReduce**
 - Mapper assigned to each block of data
 - Key-value pairs are both the input to and output of each phase
 - Keys must implement WritableComparable interface
 - Shuffle and Sort plays a key role in solving problem