Jeffrey Aven

Sams Teach Yourself

Hadoop



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Contents at a Glance

	Preface	xii
	About the Author	xiv
	Acknowledgments	XV
Part I: Get	ting Started with Hadoop	
HOUR 1	Introducing Hadoop	1
2	Understanding the Hadoop Cluster Architecture	11
3	Deploying Hadoop	23
4	Understanding the Hadoop Distributed File System (HDFS)	45
5	Getting Data into Hadoop	63
6	Understanding Data Processing in Hadoop	81
Part II: Us	ing Hadoop	
HOUR 7	Programming MapReduce Applications	103
8	Analyzing Data in HDFS Using Apache Pig	125
9	Using Advanced Pig	141
10	Analyzing Data Using Apache Hive	165
11	Using Advanced Hive	185
12	Using SQL-on-Hadoop Solutions	207
13	Introducing Apache Spark	221
14	Using the Hadoop User Environment (HUE)	243
15	Introducing NoSQL	263
Part III: M	anaging Hadoop	
HOUR 16	Managing YARN	279
17	Working with the Hadoop Ecosystem	301
	Using Cluster Management Utilities	
19	Scaling Hadoop	355

20	Understanding Cluster Configuration	367
21	Understanding Advanced HDFS	387
22	Securing Hadoop	405
23	Administering, Monitoring, and Troubleshooting Hadoop	421
24	Integrating Hadoop into the Enterprise	443
	Index	453

Table of Contents

Preface About the Author Acknowledgments	
Part I: Getting Started with Hadoop	
HOUR 1: Introducing Hadoop	1
Hadoop and a Brief History of Big Data	1
Hadoop Explained	2
The Commercial Hadoop Landscape	5
Typical Hadoop Use Cases	6
Summary	8
Q&A	8
Workshop	9
HOUR 2: Understanding the Hadoop Cluster Architecture	11
HDFS Cluster Processes	11
YARN Cluster Processes	14
Hadoop Cluster Architecture and Deployment Modes	17
Summary	20
Q&A	21
Workshop	21
HOUR 3: Deploying Hadoop	23
Installation Platforms and Prerequisites	23
Installing Hadoop	26
Deploying Hadoop in the Cloud	38
Summary	41
Q&A	42
Workshop	42

HOUR 4: Understanding the Hadoop Distributed File System (HDFS)	45
HDFS Overview	45
Review of the HDFS Roles	48
NameNode Metadata	52
SecondaryNameNode Role	55
Interacting with HDFS	56
Summary	59
Q&A	60
Workshop	60
HOUR 5: Getting Data into Hadoop	63
Data Ingestion Using Apache Flume	63
Ingesting Data from a Database using Sqoop	70
Data Ingestion Using HDFS RESTful Interfaces	74
Data Ingestion Considerations	77
Summary	78
Q&A	79
Workshop	79
HOUR 6: Understanding Data Processing in Hadoop	81
Introduction to MapReduce	81
MapReduce Explained	83
Word Count: The "Hello, World" of MapReduce	92
MapReduce in Hadoop	95
Summary	99
Q&A	100
Workshop	100
Part II: Using Hadoop	
HOUR 7: Programming MapReduce Applications	103
Introducing the Java MapReduce API	103
Writing a MapReduce Program in Java	109

	Advanced MapReduce API Concepts	117
	Using the MapReduce Streaming API	120
	Summary	122
	Q&A	123
	Workshop	123
HC	DUR 8: Analyzing Data in HDFS Using Apache Pig	125
	Introducing Pig	125
	Pig Latin Basics	126
	Loading Data into Pig	131
	Filtering, Projecting, and Sorting Data using Pig	133
	Built-in Functions in Pig	136
	Summary	139
	Q&A	139
	Workshop	140
HC	DUR 9: Using Advanced Pig	141
	Grouping Data in Pig	141
	Multiple Dataset Processing in Pig	144
	User-Defined Functions in Pig	156
	Automating Pig Using Macros and Variables	159
	Summary	161
	Q&A	162
	Workshop	162
нС	OUR 10: Analyzing Data Using Apache Hive	165
	Introducing Hive	165
	Creating Hive Objects	171
	Analyzing Data with Hive	175
	Data Output with Hive	180
	Summary	181
	Q&A	181
	Workshop	182

HOUR 11: Using Advanced Hive	185
Automating Hive	185
Complex Datatypes in Hive	187
Text Processing Using Hive	190
Optimizing and Managing Queries in Hive	199
Summary	204
Q&A	204
Workshop	205
HOUR 12: Using SQL-on-Hadoop Solutions	207
What Is SQL on Hadoop?	207
Columnar Storage in Hadoop	208
Introduction to Impala	211
Introduction to Tez	214
Introduction to HAWQ and Drill	216
Summary	217
Q&A	218
Workshop	218
HOUR 13: Introducing Apache Spark	221
Introducing Spark	221
Spark Architecture	225
Resilient Distributed Datasets in Spark	227
Transformations and Actions in Spark	231
Extensions to Spark	234
Summary	240
Q&A	241
Workshop	241
HOUR 14: Using the Hadoop User Environment (HUE)	243
Introducing HUE	243
Installing, Configuring, and Using HUE	251
Summary	260

Q&A	260
Workshop	261
HOUR 15: Introducing NoSQL	263
Introduction to NoSQL	263
Introducing HBase	265
Introducing Apache Cassandra	273
Other NoSQL Implementations and the Future of NoSQL	275
Summary	276
Q&A	276
Workshop	277
Part III: Managing Hadoop	
HOUR 16: Managing YARN	279
YARN Revisited	279
Administering YARN	285
Application Scheduling in YARN	293
Summary	299
Q&A	299
Workshop	300
HOUR 17: Working with the Hadoop Ecosystem	301
Hadoop Ecosystem Overview	301
Introduction to Oozie	306
Stream Processing and Messaging in Hadoop	311
Infrastructure and Security Projects	315
Machine Learning, Visualization, and More Data Analysis Tools	319
Summary	325
Q&A	326
Workshop	326
HOUR 18: Using Cluster Management Utilities	329
Cluster Management Overview	329
Deploying Clusters and Services Using Management Tools	337

Configuration and Service Management Using Managemen	
Monitoring, Troubleshooting, and Securing Hadoop Clusters	
Using Cluster Management Utilities	
Getting Started with the Cluster Management Utilities	
Summary	
Q&A	
Workshop	353
HOUR 19: Scaling Hadoop	355
Linear Scalability with Hadoop	355
Adding Nodes to your Hadoop Cluster	356
Decommissioning Nodes from your Cluster	359
Rebalancing a Hadoop Cluster	361
Benchmarking Hadoop	362
Summary	365
Q&A	365
Workshop	366
HOUR 20: Understanding Cluster Configuration	367
Configuration in Hadoop	367
HDFS Configuration Parameters	372
YARN Configuration Parameters	377
Ecosystem Component Configuration	381
Summary	383
Q&A	384
Workshop	385
HOUR 21: Understanding Advanced HDFS	387
HDFS Rack Awareness	387
HDFS High Availability	390
HDFS Federation	
HDFS Caching, Snapshotting, and Archiving	401
Summary	
Q&A	
Workshop	404

HOUR 22: Securing Hadoop	405
Hadoop Security Basics	405
Securing Hadoop with Kerberos	411
Perimeter Security Using Apache Knox	414
Role-Based Access Control Using Ranger and Sentry	416
Summary	418
Q&A	419
Workshop	419
HOUR 23: Administering, Monitoring, and Troubleshooting Hadoop	421
Administering Hadoop	421
Troubleshooting Hadoop	426
System and Application Monitoring in Hadoop	432
Best Practices and Other Information Sources	439
Summary	441
Q&A	441
Workshop	442
HOUR 24: Integrating Hadoop into the Enterprise	443
Hadoop and the Data Center	443
Use Case: Data Warehouse/ETL Offload	445
Use Case: Event Storage and Processing	447
Use Case: Predictive Analytics	448
Summary	450
Q&A	451
Workshop	451
Index	453

Preface

Hadoop is synonymous with Big Data, and the two are inexorably linked together. Although there have been many books written about Hadoop before this one, many of these books have been focused on one particular area of Hadoop, or required some prior experience with Hadoop. This book is different as it explores all areas of Hadoop and the Hadoop ecosystem, as well as providing an understanding and background to the genesis of Hadoop and the big data movement.

This book is also useful if you have had some exposure to Hadoop as it explores adjacent technologies such as Spark, HBase, Cassandra, and more. The book includes many diagrams, code snippets, hands-on exercises, quizzes, and Q&As, which will help you in distilling key concepts.

I have dedicated the past several years of my career to this subject area, teaching courses and consulting to clients on analytics and big data. I have seen the emergence and maturity of the big data and open source movements, and been part of its assimilation into the enterprise. I have tried to synthesize my personal learning journey over this time into this book.

I hope this book launches or assists in your journey to becoming a big data practitioner.

How This Book Is Organized

This book starts by establishing some of the basic concepts behind Hadoop, which are covered in **Part I**, "**Getting Started with Hadoop**." I also cover deployment of Hadoop both locally and in the cloud in Part I.

Part II, "Using Hadoop," is focused on the programming and data interfaces available with Hadoop, which include MapReduce, Pig, Hive, Spark, and more, as well as introductions to SQL-on-Hadoop and NoSQL using HBase.

Part III, "Managing Hadoop," covers scaling, management, and administration of Hadoop and its various related technologies, including advanced configuration, securing, monitoring, and troubleshooting Hadoop.

Data Used in the Exercises

Data for the Try it Yourself exercises can be downloaded from the book's Amazon Web Services (AWS) S3 bucket. If you are not familiar with AWS, don't worry—I cover this topic in the book as well.

Conventions Used in This Book

Each hour begins with "What You'll Learn in This Hour," which provides a list of bullet points highlighting the topics covered in that hour. Each hour concludes with a "Summary" page summarizing the main points covered in the hour, as well as "Q&A" and "Quiz" sections to help you consolidate your learning from that hour.

Key topics being introduced for the first time are typically *italicized* by convention. Most hours also include programming examples in numbered code listings. Where functions, commands, classes, or objects are referred to in text, they appear in monospace type.

Other asides in this book include the following:

NOTE
Content not integral to the subject matter but worth noting or being aware of.
TIP
TIP Subtitle
A hint or tip relating to the current topic that could be useful.
CAUTION
Caution Subtitle

TRY IT YOURSELF



Exercise Title

An exercise related to the current topic including a step-by-step guide and descriptions of expected outputs.

Something related to the current topic that could lead to issues if not addressed.

About the Author

Jeffrey Aven is a big data, open source software, and cloud computing consultant and instructor based in Melbourne, Australia. Jeff has several years' experience with Hadoop, NoSQL, Spark, and related technologies, and has been involved in key roles with several major big data implementations in Australia. Jeffrey is the author of SAMS Teach Yourself Apache Spark and was awarded Cloudera Hadoop Instructor of the Year for APAC in 2013.

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We Want to Hear from You

As the reader of this book, *you* are our most important critic and commentator. We value your opinion and want to know what we're doing right, what we could do better, what areas you'd like to see us publish in, and any other words of wisdom you're willing to pass our way.

We welcome your comments. You can email or write to let us know what you did or didn't like about this book—as well as what we can do to make our books better.

Please note that we cannot help you with technical problems related to the topic of this book.

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HOUR 3 **Deploying Hadoop**

What You'll Learn in This Hour:

- ▶ Installation platforms and prerequisites
- ▶ How to install Hadoop from Apache releases
- ▶ How to deploy Hadoop using commercial distributions
- ▶ How to deploy Hadoop in the cloud using AWS

Now that you have been introduced to Hadoop and learned about its core components, HDFS and YARN and their related processes, as well as different deployment modes for Hadoop, let's look at the different options for getting a functioning Hadoop cluster up and running. By the end of this hour you will have set up a working Hadoop cluster that we will use throughout the remainder of the book.

Installation Platforms and Prerequisites

Before you install Hadoop there are a few installation requirements, prerequisites, and recommendations of which you should be aware.

Operating System Requirements

The vast majority of Hadoop implementations are platformed on Linux hosts. This is due to a number of reasons:

- ▶ The Hadoop project, although cross-platform in principle, was originally targeted at Linux. It was several years after the initial release that a Windows-compatible distribution was introduced.
- ▶ Many of the commercial vendors only support Linux.
- ▶ Many other projects in the open source and Hadoop ecosystem have compatibility issues with non-Linux platforms.

That said there are options for installing Hadoop on Windows, should this be your platform of choice. We will use Linux for all of our exercises and examples in this book, but consult the documentation for your preferred Hadoop distribution for Windows installation and support information if required.

If you are using Linux, choose a distribution you are comfortable with. All major distributions are supported (Red Hat, Centos, Ubuntu, SLES, etc.). You can even mix distributions if appropriate; for instance, master nodes running Red Hat and slave nodes running Ubuntu.

CAUTION

Don't Use Logical Volume Manager (LVM) in Linux

If you are using Linux to deploy Hadoop nodes, master or slaves, it is strongly recommended that you not use LVM in Linux. This will restrict performance, especially on slave nodes.

Hardware Requirements

Although there are no hard and fast requirements, there are some general heuristics used in sizing instances, or hosts, appropriately for roles within a Hadoop cluster. First, you need to distinguish between master and slave node instances, and their requirements.

Master Nodes

A Hadoop cluster relies on its master nodes, which host the NameNode and ResourceManager, to operate, although you can implement high availability for each subsystem as I discussed in the last hour. Failure and failover of these components is not desired. Furthermore, the master node processes, particularly the NameNode, require a large amount of memory to operate efficiently, as you will appreciate in the next hour when we dive into the internals of HDFS. Therefore, when specifying hardware requirements the following guidelines can be used for medium to large-scale production Hadoop implementations:

- ▶ 16 or more CPU cores (preferably 32)
- ▶ 128GB or more RAM (preferably 256GB)
- ▶ RAID Hard Drive Configuration (preferably with hot-swappable drives)
- Redundant power supplies
- ▶ Bonded Gigabit Ethernet or 10Gigabit Ethernet

This is only a guide, and as technology moves on quickly, these recommendations will change as well. The bottom line is that you need *carrier class* hardware with as much CPU and memory capacity as you can get!

Slave Nodes

Slave nodes do the actual work in Hadoop, both for processing and storage so they will benefit from more CPU and memory—physical memory, not virtual memory. That said, slave nodes are designed with the expectation of failure, which is one of the reasons blocks are replicated in HDFS. Slave nodes can also be scaled out linearly. For instance, you can simply add more nodes to add more aggregate storage or processing capacity to the cluster, which you cannot do with master nodes. With this in mind, economic scalability is the objective when it comes to slave nodes. The following is a guide for slave nodes for a well-subscribed, computationally intensive Hadoop cluster; for instance, a cluster hosting machine learning and in memory processing using Spark.

- ▶ 16-32 CPU cores
- ▶ 64-512 GB of RAM
- ▶ 12-24 1-4 TB hard disks in a JBOD Configuration

NOTE

JBOD

JBOD is an acronym for *just a bunch of disks*, meaning directly attached storage that is not in a RAID configuration, where each disk operates independently of the other disks. RAID is not recommended for block storage on slave nodes as the access speed is limited by the slowest disk in the array, unlike JBOD where the average speed can be greater than that of the slowest disk. JBOD has been proven to outperform RAID 0 for block storage by 30% to 50% in benchmarks conducted at Yahoo!.

CAUTION

Storing Too Much Data on Any Slave Node May Cause Issues

As slave nodes typically host the blocks in a Hadoop filesystem, and as storage costs, particularly for JBOD configurations, are relatively inexpensive, it may be tempting to allocate excess block storage capacity to each slave node. However, as you will learn in the next hour on HDFS, you need to consider the network impact of a failed node, which will trigger re-replication of all blocks that were stored on the slave node.

Slave nodes are designed to be deployed on commodity-class hardware, and yet while they still need ample processing power in the form of CPU cores and memory, as they will be executing computational and data transformation tasks, they don't require the same degree of fault tolerance that master nodes do.

Networking Considerations

Fully distributed Hadoop clusters are very chatty, with control messages, status updates and heartbeats, block reports, data shuffling, and block replication, and there is often heavy network utilization between nodes of the cluster. If you are deploying Hadoop on-premise, you should always deploy Hadoop clusters on private subnets with dedicated switches. If you are using multiple racks for your Hadoop cluster (you will learn more about this in Hour 21, "Understanding Advanced HDFS"), you should consider redundant core and "top of rack" switches.

Hostname resolution is essential between nodes of a Hadoop cluster, so both forward and reverse DNS lookups must work correctly between each node (master-slave and slave-slave) for Hadoop to function. Either DNS or a hosts files can be used for resolution. IPv6 should also be disabled on all hosts in the cluster.

Time synchronization between nodes of the cluster is essential as well, as some components, such as Kerberos, which is discussed in **Hour 22**, "Securing Hadoop," rely on this being the case. It is recommended you use ntp (Network Time Protocol) to keep clocks synchronized between all nodes.

Software Requirements

As discussed, Hadoop is almost entirely written in Java and compiled to run in a Java Runtime Environment (JRE); therefore Java is a prerequisite to installing Hadoop. Current prerequisites include:

- ▶ Java Runtime Envrionment (JRE) 1.7 or above
- ▶ Java Development Kit (JDK) 1.7 or above—required if you will be compiling Java classes such as MapReduce applications

Other ecosystem projects will have their specific prerequisites; for instance, Apache Spark requires Scala and Python as well, so you should always refer to the documentation for these specific projects.

Installing Hadoop

You have numerous options for installing Hadoop and setting up Hadoop clusters. As Hadoop is a top-level Apache Software Foundation (ASF) open source project, one method is to install directly from the Apache builds on http://hadoop.apache.org/. To do this you first need one or more hosts, depending upon the mode you wish to use, with appropriate hardware specifications, an appropriate operating system, and a Java runtime environment available (all of the prerequisites and considerations discussed in the previous section).

Once you have this, it is simply a matter of downloading and unpacking the desired release. There may be some additional configuration to be done afterwards, but then you simply start the relevant services (master and slave node daemons) on their designated hosts and you are up and running.

Non-Commercial Hadoop

Let's deploy a Hadoop cluster using the latest Apache release now.

TRY IT YOURSELF



Installing Hadoop Using the Apache Release

In this exercise we will install a pseudo-distributed mode Hadoop cluster using the latest Hadoop release downloaded from hadoop.apache.org.

As this is a test cluster the following specifications will be used in our example:

- ▶ Red Hat Enterprise Linux 7.2 (The installation steps would be similar using other Linux distributions such as Ubuntu)
- ▶ 2 CPU cores
- ▶ 8GB RAM
- ▶ 30GB HDD
- ▶ hostname: hadoopnode0
- 1. Disable SELinux (this is known to cause issues with Hadoop):

```
$ sudo sed -i 's/SELINUX=enforcing/SELINUX=disabled/g' \
/etc/selinux/config
```

2. Disable IPv6 (this is also known to cause issues with Hadoop):

```
$ sudo sed -i "\$anet.ipv6.conf.all.disable_ipv6 = 1" \
/etc/sysctl.conf
$ sudo sed -i "\$anet.ipv6.conf.default.disable_ipv6 = 1" \
/etc/sysctl.conf
$ sudo sysctl -p
```

- 3. Reboot
- **4.** Run the sestatus command to ensure SELinux is not enabled:

```
$ sestatus
```



- 5. Install Java. We will install the OpenJDK, which will install both a JDK and JRE:
 - \$ sudo yum install java-1.7.0-openjdk-devel
 - a. Test that Java has been successfully installed by running the following command:

```
$ java -version
```

If Java has been installed correctly you should see output similar to the following:

```
java version "1.7.0_101"
OpenJDK Runtime Environment (rhel-2.6.6.1.el7_2-x86_64..)
OpenJDK 64-Bit Server VM (build 24.95-b01, mixed mode)
```

Note that depending upon which operating system you are deploying on, you may have a version of Java and a JDK installed already. In these cases it may not be necessary to install the JDK, or you may need to set up alternatives so you do not have conflicting Java versions.

6. Locate the installation path for Java, and set the JAVA HOME environment variable:

```
$ export JAVA_HOME=/usr/lib/jvm/REPLACE_WITH_YOUR_PATH/
```

7. Download Hadoop from your nearest Apache download mirror. You can obtain the link by selecting the binary option for the version of your choice at http://hadoop.apache.org/releases.html. We will use Hadoop version 2.7.2 for our example.

```
$ wget http://REPLACE_WITH_YOUR_MIRROR/hadoop-2.7.2.tar.gz
```

8. Unpack the Hadoop release, move it into a system directory, and set an environment variable from the Hadoop home directory:

```
$ tar -xvf hadoop-2.7.2.tar.gz
$ mv hadoop-2.7.2 hadoop
$ sudo mv hadoop/ /usr/share/
$ export HADOOP_HOME=/usr/share/hadoop
```

9. Create a directory which we will use as an alternative to the Hadoop configuration directory:

```
$ sudo mkdir -p /etc/hadoop/conf
```

10. Create a mapred-site.xml file (I will discuss this later) in the Hadoop configuration directory:

```
$ sudo cp $HADOOP_HOME/etc/hadoop/mapred-site.xml.template \
$HADOOP HOME/etc/hadoop/mapred-site.xml
```

11. Add JAVA_HOME environment variable to hadoop-env.sh (file used to source environment variables for Hadoop processes):



```
$ sed -i "\$aexport JAVA_HOME=/REPLACE_WITH_YOUR_JDK_PATH/" \
$HADOOP HOME/etc/hadoop/hadoop-env.sh
```

Substitute the correct path to your Java home directory as defined in Step 6.

12. Create a symbolic link between the Hadoop configuration directory and the /etc/hadoop /conf directory created in Step 10:

```
$ sudo ln -s $HADOOP_HOME/etc/hadoop/* \
/etc/hadoop/conf/
```

13. Create a logs directory for Hadoop:

```
$ mkdir $HADOOP HOME/logs
```

14. Create users and groups for HDFS and YARN:

```
$ sudo groupadd hadoop
$ sudo useradd -g hadoop hdfs
$ sudo useradd -g hadoop yarn
```

15. Change the group and permissions for the Hadoop release files:

```
$ sudo chgrp -R hadoop /usr/share/hadoop
$ sudo chmod -R 777 /usr/share/hadoop
```

16. Run the built in Pi Estimator example included with the Hadoop release.

```
$ cd $HADOOP_HOME
$ sudo -u hdfs bin/hadoop jar \
share/hadoop/mapreduce/hadoop-mapreduce-examples-2.7.2.jar \
pi 16 1000
```

As we have not started any daemons or initialized HDFS, this program runs in LocalJobRunner mode (recall that I discussed this in **Hour 2**, "Understanding the Hadoop Cluster Architecture"). If this runs correctly you should see output similar to the following:

Now let's configure a pseudo-distributed mode Hadoop cluster from your installation.



17. Use the vi editor to update the core-site.xml file, which contains important information about the cluster, specifically the location of the namenode:

```
$ sudo vi /etc/hadoop/conf/core-site.xml
# add the following config between the <configuration>
# and </configuration> tags:
property>
<name>fs.defaultFS</name>
<value>hdfs://hadoopnode0:8020</value>
</property>
```

Note that the value for the fs.defaultFS configuration parameter needs to be set to hdfs://HOSTNAME:8020, where the HOSTNAME is the name of the NameNode host, which happens to be the localhost in this case.

18. Adapt the instructions in Step 17 to similarly update the hdfs-site.xml file, which contains information specific to HDFS, including the replication factor, which is set to 1 in this case as it is a pseudo-distributed mode cluster:

```
sudo vi /etc/hadoop/conf/hdfs-site.xml
# add the following config between the <configuration>
# and </configuration> tags:
property>
<name>dfs.replication</name>
<value>1</value>
</property>
```

19. Adapt the instructions in Step 17 to similarly update the yarn-site.xml file, which contains information specific to YARN. Importantly, this configuration file contains the address of the resourcemanager for the cluster—in this case it happens to be the localhost, as we are using pseudo-distributed mode:

```
$ sudo vi /etc/hadoop/conf/yarn-site.xml
# add the following config between the <configuration>
# and </configuration> tags:
property>
<name>yarn.resourcemanager.hostname</name>
<value>hadoopnode0</value>
</property>
property>
<name>yarn.nodemanager.aux-services</name>
<value>mapreduce shuffle</value>
</property>
```

20. Adapt the instructions in Step 17 to similarly update the mapred-site.xml file, which contains information specific to running MapReduce applications using YARN:

```
$ sudo vi /etc/hadoop/conf/mapred-site.xml
# add the following config between the <configuration>
```

```
# and </configuration> tags:
<name>mapreduce.framework.name</name>
<value>yarn</value>
```



21. Format HDFS on the NameNode:

```
$ sudo -u hdfs bin/hdfs namenode -format
```

Enter [Y] to re-format if prompted.

22. Start the NameNode and DataNode (HDFS) daemons:

```
$ sudo -u hdfs sbin/hadoop-daemon.sh start namenode
$ sudo -u hdfs sbin/hadoop-daemon.sh start datanode
```

23. Start the ResourceManager and NodeManager (YARN) daemons:

```
$ sudo -u yarn sbin/yarn-daemon.sh start resourcemanager
$ sudo -u yarn sbin/yarn-daemon.sh start nodemanager
```

24. Use the jps command included with the Java JDK to see the Java processes that are running:

```
$ sudo jps
```

You should see output similar to the following:

```
2374 DataNode2835 Jps2280 NameNode2485 ResourceManager2737 NodeManager
```

25. Create user directories and a tmp directory in HDFS and set the appropriate permissions and ownership:

```
$ sudo -u hdfs bin/hadoop fs -mkdir -p /user/<your_user>
$ sudo -u hdfs bin/hadoop fs -chown <your_user>:<your_user> /user/<your_user>
$ sudo -u hdfs bin/hadoop fs -mkdir /tmp
$ sudo -u hdfs bin/hadoop fs -chmod 777 /tmp
```

26. Now run the same Pi Estimator example you ran in Step 16. This will now run in pseudo-distributed mode:

```
$ bin/hadoop jar \
share/hadoop/mapreduce/hadoop-mapreduce-examples-2.7.2.jar \
pi 16 1000
```



The output you will see in the console will be similar to that in Step 16. Open a browser and go to localhost:8088. You will see the YARN ResourceManager Web UI (which I discuss in **Hour 6**, "Understanding Data Processing in Hadoop") (Figure 3.1):

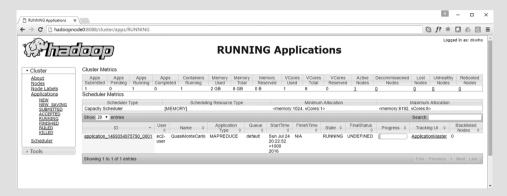


FIGURE 3.1 YARN ResourceManager Web UI.

Congratulations! You have just set up your first pseudo-distributed mode Hadoop cluster. We will use this cluster in other exercises in the book, so keep it available if you can.

Using a Commercial Hadoop Distribution

As I had discussed in **Hour 1**, "**Introducing Hadoop**," the commercial Hadoop landscape is well established. With the advent of the ODPi (the Open Data Platform initiative), a once-numerous array of vendors and derivative distributions has been consolidated to a much simpler landscape which includes three primary *pure-play* Hadoop vendors:

- ▶ Cloudera
- ▶ Hortonworks
- ▶ MapR

Importantly, enterprise support agreements and subscriptions can be purchased from the various Hadoop vendors for their distributions. Each vendor also supplies a suite of management utilities to help you deploy and manage Hadoop clusters. Let's look at each of the three major pure play Hadoop vendors and their respective distributions.

Cloudera

Cloudera was the first mover in the commercial Hadoop space, establishing their first commercial release in 2008. Cloudera provides a Hadoop distribution called CDH (Cloudera Distribution of Hadoop), which includes the Hadoop core and many ecosystem projects. CDH is entirely open source.

Cloudera also provides a management utility called Cloudera Manager (which is not open source). Cloudera Manager provides a management console and framework enabling the deployment, management, and monitoring of Hadoop clusters, and which makes many administrative tasks such as setting up high availability or security much easier. The mix of open source and proprietary software is often referred to as **open core**. A screenshot showing Cloudera Manager is pictured in Figure 3.2.

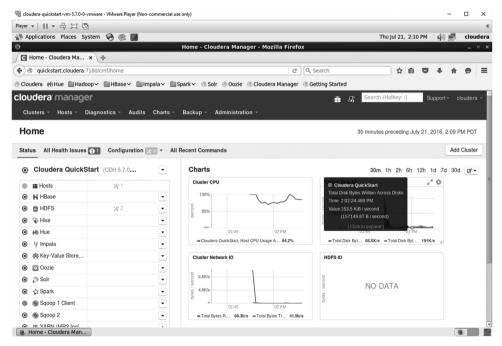


FIGURE 3.2 Cloudera Manager.

As mentioned, Cloudera Manager can be used to deploy Hadoop clusters, including master nodes, slave nodes, and ecosystem technologies. Cloudera Manager distributes installation packages for Hadoop components through a mechanism called *parcels*. As Hadoop installations are typically isolated from public networks, Cloudera Manager, which is technically not part of the cluster and will often have access to the Internet, will download parcels and distribute

these to new target hosts nominated to perform roles in a Hadoop cluster or to existing hosts to upgrade components.

Deploying a fully distributed CDH cluster using Cloudera Manager would involve the following steps at a high level:

- Install Cloudera Manager on a host that has access to other hosts targeted for roles in the cluster.
- **2.** Specify target hosts for the cluster using Cloudera Manager.
- **3.** Use Cloudera Manager to provision Hadoop services, including master and slave nodes.

Cloudera also provides a Quickstart virtual machine, which is a pre-configured pseudo-distributed Hadoop cluster with the entire CDH stack, including core and ecosystem components, as well as a Cloudera Manager installation. This virtual machine is available for VirtualBox, VMware, and KVM, and works with the free editions of each of the virtualization platforms. The Cloudera Quickstart VM is pictured in Figure 3.3.

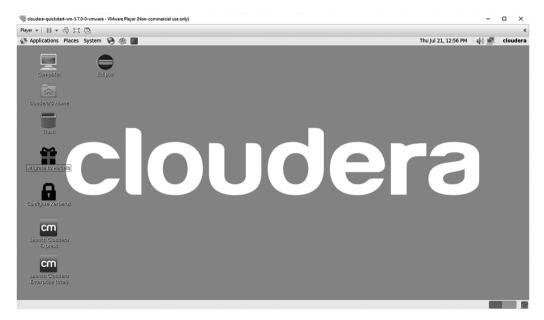


FIGURE 3.3 Cloudera Quickstart VM.

The Quickstart VM is a great way to get started with the Cloudera Hadoop offering. To find out more, go to http://www.cloudera.com/downloads.html.

More information about Cloudera is available at http://www.cloudera.com/.

Hortonworks

Hortonworks provides pure open source Hadoop distribution and a founding member of the open data platform initiative (ODPi) discussed in **Hour 1**. Hortonworks delivers a distribution called HDP (Hortonworks Data Platform), which is a complete Hadoop stack including the Hadoop core and selected ecosystem components. Hortonworks uses the Apache Ambari project to provide its deployment configuration management and cluster monitoring facilities. A screenshot of Ambari is shown in Figure 3.4.

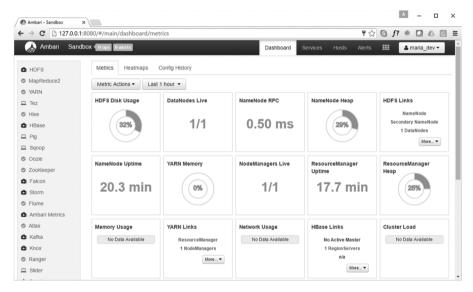


FIGURE 3.4 Ambari console.

The simplest method to deploy a Hortonworks Hadoop cluster would involve the following steps:

- **1.** Install Ambari using the Hortonworks installer on a selected host.
- **2.** Add hosts to the cluster using Ambari.
- **3.** Deploy Hadoop services (such as HDFS and YARN) using Ambari.

Hortonworks provides a fully functional, pseudo-distributed HDP cluster with the complete Hortonworks application stack in a virtual machine called the Hortonworks Sandbox. The Hortonworks Sandbox is available for VirtualBox, VMware, and KVM. The Sandbox virtual machine includes several demo applications and learning tools to use to explore Hadoop and its various projects and components. The Hortonworks Sandbox welcome screen is shown in Figure 3.5.

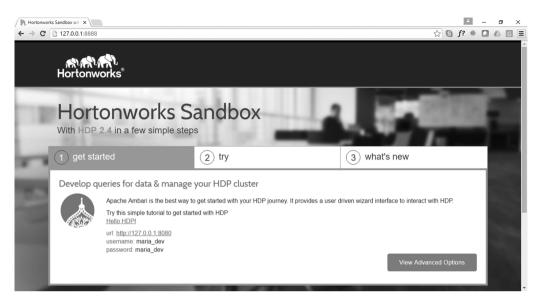


FIGURE 3.5 Hortonworks Sandbox.

You can download the Hortonworks Sandbox from http://hortonworks.com/products/sandbox/. More information about Hortonworks and HDP can be obtained from http://hortonworks.com/.

MapR

MapR delivers a "Hadoop-derived" software platform that implements an API-compatible distributed filesystem called MapRFS (the MapR distributed Filesystem). MapRFS has been designed to maximize performance and provide read-write capabilities not offered by native HDFS. MapR delivers three versions of their offering called the "Converged Data Platform." These include:

- ▶ M3 or "Converged Community Edition" (free version)
- ▶ M5 or "Converged Enterprise Edition" (supported version)
- ▶ M7 (M5 version that includes MapR's custom HBase-derived data store)

Like the other distributions, MapR has a demo offering called the "MapR Sandbox," which is available for VirtualBox or VMware. It is pictured in Figure 3.6.

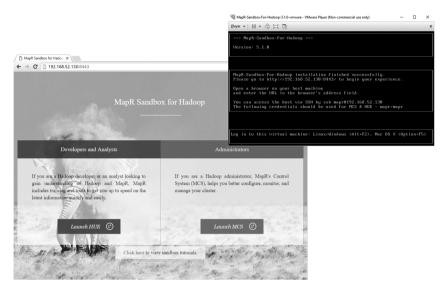


FIGURE 3.6 MapR Sandbox VM.

The MapR Sandbox can be downloaded from https://www.mapr.com/products/mapr-sandbox-hadoop/download.

MapR's management offering is called the MapR Control System (MCS), which offers a central console to configure, monitor and manage MapR clusters. It is shown in Figure 3.7.

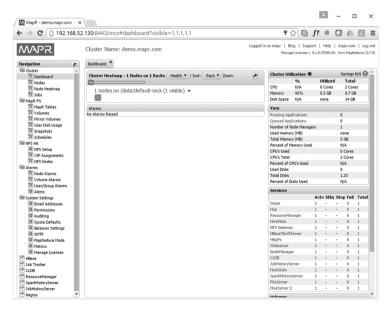


FIGURE 3.7 MapR Control System (MCS).

Much more information about MapR and the Converged Data Platform is available at https://www.mapr.com/.

Deploying Hadoop in the Cloud

The rise and proliferation of cloud computing and virtualization technologies has definitely been a game changer for the way organizations think about and deploy technology, and Hadoop is no exception. The availability and maturity around IaaS (Infrastructure-as-a-Service), PaaS (Platform-as-a-Service) and SaaS (Software-as-a-Service) solutions makes deploying Hadoop in the cloud not only viable but, in some cases, desirable.

There are many public cloud variants that can be used to deploy Hadoop including Google, IBM, Rackspace, and others. Perhaps the most pervasive cloud platform to date has been AWS (Amazon Web Services), which I will use as the basis for our discussions.

Before you learn about deployment options for Hadoop in AWS, let's go through a quick primer and background on some of the key AWS components. If you are familiar with AWS, feel free to jump straight to the Try it Yourself exercise on deploying Hadoop using AWS EMR.

EC₂

Elastic Compute Cloud (EC2) EC2 is Amazon's web service-enabled virtual computing platform. EC2 enables users to create virtual servers and networks in the cloud. The virtual servers are called instances. EC2 instances can be created with a variety of different instance permutations. The Instance Type property determines the number of virtual CPUs and the amount of memory and storage an EC2 instance has available to it. An example instance type is m4.large. A complete list of the different EC2 instance types available can be found at https://aws.amazon.com/ec2/instance-types/.

EC2 instances can be optimized for compute, memory, storage and mixed purposes and can even include GPUs (Graphics Processing Units), a popular option for machine learning and deep analytics.

There are numerous options for operating systems with EC2 instances. All of the popular Linux distributions are supported, including Red Hat, Ubuntu, and SLES, as well various Microsoft Windows options.

EC2 instances are created in security groups. Security groups govern network permissions and Access Control Lists (ACLs). Instances can also be created in a Virtual Private Cloud (VPC). A VPC is a private network, not exposed directly to the Internet. This is a popular option for organizations looking to minimize exposure of EC2 instances to the public Internet.

EC2 instances can be provisioned with various storage options, including instance storage or ephemeral storage, which are terms for volatile storage that is lost when an instance is stopped,

and Elastic Block Store (EBS), which is persistent, fault-tolerant storage. There are different options with each, such as SSD (solid state) for instance storage, or provisioned IOPS with EBS.

Additionally, AWS offers Spot instances, which enable you to bid on spare Amazon EC2 computing capacity, often available at a discount compared to normal on-demand EC2 instance pricing.

EC2, as well as all other AWS services, is located in an AWS region. There are currently nine regions, which include the following:

- ▶ US East (N. Virginia)
- ▶ US West (Oregon)
- ▶ US West (N. California)
- ▶ EU (Ireland)
- ► EU (Frankfurt)
- ► Asia Pacific (Singapore)
- ► Asia Pacific (Sydney)
- ► Asia Pacific (Tokyo)
- ▶ South America (Sao Paulo)

S3

Simple Storage Service (S3) is Amazon's cloud-based object store. An object store manages data (such as files) as objects. These objects exist in buckets. Buckets are logical, user-created containers with properties and permissions. S3 provides APIs for users to create and manage buckets as well as to create, read, and delete objects from buckets.

The S3 bucket namespace is global, meaning that any buckets created must have a globally unique name. The AWS console or APIs will let you know if you are trying to create a bucket with a name that already exists.

S3 objects, like files in HDFS, are immutable, meaning they are write once, read many. When an S3 object is created and uploaded, an ETag is created, which is effectively a signature for the object. This can be used to ensure integrity when the object is accessed (downloaded) in the future.

There are also public buckets in S3 containing public data sets. These are datasets provided for informational or educational purposes, but they can be used for data operations such as processing with Hadoop. Public datasets, many of which are in the tens or hundreds of terabytes, are available, and topics range from historical weather data to census data, and from astronomical data to genetic data.

More information about S3 is available at https://aws.amazon.com/s3/.

Elastic MapReduce (EMR)

Elastic MapReduce (EMR) is Amazon's Hadoop-as-a-Service platform. EMR clusters can be provisioned using the AWS Management Console or via the AWS APIs. Options for creating EMR clusters include number of nodes, node instance types, Hadoop distribution, and additional ecosystem applications to install.

EMR clusters can read data and output results directly to and from S3. They are intended to be provisioned on demand, run a discrete workflow, a job flow, and terminate. They do have local storage, but they are not intended to run in perpetuity. You should only use this local storage for transient data.

EMR is a quick and scalable deployment method for Hadoop. More information about EMR can be found at https://aws.amazon.com/elasticmapreduce/.

AWS Pricing and Getting Started

AWS products, including EC2, S3, and EMR, are charged based upon usage. Each EC2 instance type within each region has an instance per hour cost associated with it. The usage costs per hour are usually relatively low and the medium- to long-term costs are quite reasonable, but the more resources you use for a longer period of time, the more you are charged.

If you have not already signed up with AWS, you're in luck! AWS has a free tier available for new accounts that enables you to use certain instance types and services for free for the first year. You can find out more at https://aws.amazon.com/free/. This page walks you through setting up an account with no ongoing obligations.

Once you are up and running with AWS, you can create an EMR cluster by navigating to the EMR link in the AWS console as shown in Figure 3.8.



FIGURE 3.8
AWS console—EMR option.

Then click Create Cluster on the EMR welcome page as shown in Figure 3.9, and simply follow the dialog prompts.

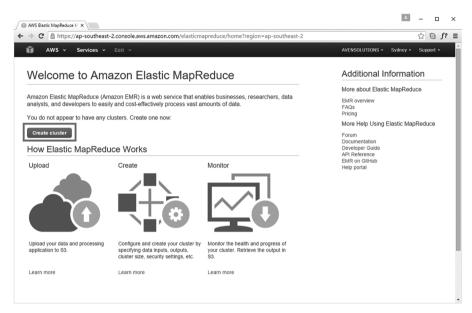


FIGURE 3.9
AWS EMR welcome screen.

You can use an EMR cluster for many of our exercises. However, be aware that leaving the cluster up and running will incur usage charges.

Summary

In this hour you learned about the various approaches to deploying a Hadoop cluster including the Apache releases, commercial distributions and cloud deployment options. Commercial distributions are often the best approach to deploying Hadoop on premise in most organizations as these distributions provide a stable, tested combination of core and ecosystem releases, as well as typically providing a suite of management capabilities useful for deploying and managing Hadoop clusters at scale.

You also learned how to provision Hadoop clusters in the cloud by using the Amazon Web Services Hadoop-as-a-Service offering—Elastic MapReduce (EMR). You are encouraged to explore all the options available to deploy Hadoop. As you progress through the book you will be performing hands-on exercises using Hadoop, so you will need to have a functional cluster. This could be one of the sandbox or quickstart commercial offerings or the Apache Hadoop cluster we set up in the Try it Yourself exercise in this hour.

Q&A

- Q. Why do master nodes normally require a higher degree of fault tolerance than slave nodes?
- **A.** Slave nodes are designed to be implemented on commodity hardware with the expectation of failure, and this enables slave nodes to scale economically. The fault tolerance and resiliency built into HDFS and YARN enables the system to recover seamlessly from a failed slave node. Master nodes are different; they are intended to be "always on." Although there are high availability implementation options for master nodes, failover is not desirable. Therefore, more local fault tolerance, such as RAID disks, dual power supplies, etc., is preferred for master nodes.
- Q. What does JBOD stand for, and what is its relevance for Hadoop?
- A. JBOD is an acronym for "Just a Bunch of Disks," which means spinning disks that operate independently of one another, in contrast to RAID, where disks operate as an array. JBOD is recommended for slave nodes, which are responsible for HDFS block storage. This is because the average speed of all disks on a slave node is greater than the speed of the slowest disk. By comparison, RAID read and write speeds are limited by the speed of the slowest disk in the array.
- Q. What are the advantages to deploying Hadoop using a commercial distribution?
- **A.** Commercial distributions contain a "stack" of core and ecosystem components that are tested with one another and certified for the respective distribution. The commercial vendors typically include a management application, which is very useful for managing multinode Hadoop clusters at scale. The commercial vendors also offer enterprise support as an option as well.

Workshop

The workshop contains quiz questions to help you solidify your understanding of the material covered. Try to answer all questions before looking at the "Answers" section that follows.

Quiz

- 1. True or False: A Java Runtime Environment (JRE) is required on hosts that run Hadoop services.
- 2. Which AWS PaaS product is used to deploy Hadoop as a service?
 - A. EC2
 - B. EMR
 - **C.** S3
 - **D.** DynamoDB

- **3.** Slave nodes are typically deployed on what class of hardware?
- **4.** The open source Hadoop cluster management utility used by Hortonworks is called _____.

Answers

- **1.** True. Hadoop services and processes are written in Java, are compiled to Java bytecode, and run in Java Virtual Machines (JVMs), and therefore require a JRE to operate.
- 2. B. Elastic MapReduce (EMR).
- 3. Commodity.
- 4. Ambari.

Index

Amazon Web Services (AWS) deploying Hadoop clusters, 38-41 access control policies, 416-418 EC2, 38-39 accumulators, 436 Accumulo, 305-306 ACLs (access control lists) in HDFS, 53-54 action nodes, 307 actions on RDDs, 231 **ADD PARTITION statement** (Hive), 200 administration of HDFS, 422-424 of Hive, 425-426 of YARN, 424-425 advanced analytics use case, 7-8 aged history offload, 445 agents (Flume), 64-65 channels, 67 sinks, 66-67 sources, 65-66 aggregating data

in Hive, 176

in Pig, 143

```
EMR, 40
   pricing, 40-41
   S3, 39
Ambari. See Apache Ambari
Ambari Views, 335-336
analyzing data in Hive, 175
Apache Ambari, 333-336
   Ambari Views, 335-336
   architecture, 334-335
   configuration management,
    345-346
   deploying services, 337-340
   installing, 335, 351
   monitoring and trouble-
    shooting clusters, 347-350
   securing clusters, 351
   service management,
    344-345
Apache Avro project, 105
Apache Cassandra. See Cassandra
Apache DataFu, 158
```

Apache Drill, 208, 216-217	Spark	parameterizing queries,
Apache Flink, 208	planning, 226	185–186
Apache Flume, 63-70	scheduling, 226	auxiliary services in YARN, 284
architecture, 64-67	support for, 222	Avro format, 105, 318-319
installing, 67–70	YARN	AWS (Amazon Web Services)
Apache HAWQ, 208, 216	killing, 288–289	deploying Hadoop clusters,
Apache Hive. See Hive	running, 96–99, 281–284	38–41
Apache Kafka, 313-315, 447	scheduling, 293–298	EC2, 38–39
Apache Knox, 415-416	architecture	EMR, 40
Apache Mahout, 320, 450	of Apache Ambari, 334-335	pricing, 40–41
Apache Parquet, 209-210	of Cloudera Manager,	S3, 39
Apache Pig. See Pig	331–332	
Apache Presto, 208	of Flume, 64-67	
Apache Ranger, 417	of HBase, 268–269	В
Apache Sentry, 418	of Kafka, 314–315	
Apache Software Foundation	of Oozie, 309	backing up NameNodes, 423
(ASF), 2	of Spark, 225	bags (Pig Latin), 128, 133
Apache Spark. See Spark	drivers, 226-227	balancer utility, 361-362
Apache Storm, 311-312	executors, 227	Beeline, 168
Apache Tez, 208, 214-216	of YARN, 279-281	Beeswax, 168
Apache Thrift, 319	archiving in HDFS, 402	benchmarking Hadoop, 362–365
Apache Zeppelin, 323-324	ARRAY datatype, 187-188	best practices
appenders (Log4j), 427	ASF (Apache Software	data ingestion, 77–78
ApplicationMaster, 16, 96,	Foundation), 2	debugging, 440
281, 283	asymmetry of MapReduce, 89-90	big data, history of, 1–2
applications	atoms (Pig Latin), 128	Bigtable, 263–264
log files within, 430–431	attribute value pairs. See key	blocks in HDFS, 12, 46-48
MapReduce	value pairs	DataNodes, 49-50
compiling and packaging,	authentication, 54, 411	distribution of, 47
114–116	in HUE, 255	failure recovery, 48
components of, 106-108	for web UIs, 407-409	replication of, 47-48
counters, 434-439	authorization, 53, 411	blueprints (Ambari), 339-340
writing in Java, 109–114	Hive, 169–171	box classes, 104
Pig	in HUE, 256	bucketing in Hive, 201–202
parameterizing, 159–160	automating Hive	built-in functions
reusing code, 160–161	custom scripts, 186-187	in Hive, 175-176

in Pig, 136

C	service management, 340–342	SecondaryNameNode, 55, 60
caching in HDFS, 401	software distribution, 332	YARN, 14-17, 21, 279-281
CapacityScheduler, 295-298	cluster management,	clusters, 4
Cassandra, 273	329-330, 351	adding nodes to, 356–359
CQL, 273-274	Apache Ambari, 333–336	decommissioning nodes from,
data model, 273	Ambari Views, 335–336	359–361
with Hadoop, 274-275	architecture, 334–335	deploying
Cassandra Query Language (CQL), 273–274	configuration management, 345–346	Apache Ambari, 337–340 in cloud, 38–41
CEP (Complex Event Processing), 447–448	deploying services, 337–340	Cloudera Manager, 33–34, 340
channels (Flume agents), 67	installing, 335, 351	commercial Hadoop,
classes (Java), naming	monitoring and trouble-	33–38, 42
conventions, 106	shooting clusters,	Hortonworks, 35-36
CLI (command line interface)	347–350	MapR, 36-38
Hive, 168–169	securing clusters, 351	non-commercial Hadoop,
YARN, 289-290	service management, 344–345	27–32
client properties for HDFS,	Cloudera Manager, 330–333 architecture, 331–332	deployment modes, 18–20
376-377		Hadoop cluster architecture, 17–18
cloud, deploying Hadoop clusters,	configuration management,	in HBase, 268–269
38-41	342–344 deploying services, 340	master-slave cluster
Cloudera, 5-6, 33-34		architecture, 12
Cloudera Enterprise, 331 Cloudera Express, 331	editions, 331	monitoring
Cloudera Impala, 208, 211–214	installing, 332–333, 351	Apache Ambari, 347–350
Cloudera Manager, 330–333	monitoring and trouble-	Cloudera Manager,
architecture, 331–332	shooting clusters,	347–350
configuration management,	347–350	counters, 434-439
342–344	securing clusters, 351	MBeans, 432-434
deploying services, 340	service management, 340–342	rebalancing, 361–362
editions, 331		securing
installing, 332–333, 351	software distribution, 332	Apache Ambari, 351
monitoring and trouble-	cluster processes HDFS, 11–14, 21	Cloudera Manager, 351
shooting clusters, 347–350	DataNodes, 49–50	ecosystem projects,
securing clusters, 351	NameNode, 48–49	315–318

starting and stopping	hadoop-metrics.properties	counters, 434-439	
services, 421-422	file, 372	counting words, 92-95	
troubleshooting	HBase, 383	CQL (Cassandra Query Language),	
Apache Ambari, 347–350	HDFS	273-274	
Cloudera Manager,	client properties, 376-377	CREATE TABLE statement (Hive),	
347–350	common properties,	172	
COGROUP statement (Pig),	372–373	cross joins in Pig, 154–155	
144-145	DataNode properties,	CROSS statement (Pig), 154-155	
cogrouping multiple datasets in	375–376	custom connectors for Sqoop, 73	
Pig, 144–145	NameNode properties, 373–374	custom scripts in Hive, 186-187	
co-locating slave nodes, 21		Cutting, Doug, 1–2	
column encryption with Hive, 197	SecondaryNameNode		
columnar storage, 208–209	properties, 374–375		
ORC format, 210–211	Hive, 169, 178–179, 381	D	
Parquet format, 209-210	HUE, 251–255	<i>D</i>	
Combiner functions of	include files, 357–359	daemons, 11	
MapReduce, 88-89, 117	Job object and, 107	in HDFS, 21	
comments in Pig Latin, 131	log4j.properties file, 371	log files for, 428–430	
commercial Hadoop distributions,	Pig, 381–382	in YARN, 21, 279–281	
5-6, 33-38, 42	precedence rules, 369-370		
common properties	Spark, 382	DAGs (Directed Acyclic Graphs), 215	
for HDFS, 372-373	YARN, 285-286	data	
for YARN, 377	common properties, 377		
community function libraries in	MapReduce properties,	aggregating	
Pig, 157-158	379–380	in Hive, 176	
compiling MapReduce	NodeManager properties,	in Pig, 143	
applications, 114-116	377–379	analyzing in Hive, 175	
complex datatypes in Hive,	YARN CapacityScheduler, 298	distribution in HBase, 268–269	
187-190	configuration management		
Complex Event Processing (CEP), 447–448	Apache Ambari, 345-346	extracting with Regex SerDe, 191	
	Cloudera Manager, 342–344	filtering in Pig, 133, 150	
concatenating multiple datasets in Pig, 153	consistency in HDFS, 54	grouping	
	containers, 15, 96	in Hive, 176	
configuration, 367-368 default values, 368-369	Context object, 112-113	in Pig, 141–142	
	CONTEXT_NGRAMS function, 195	<u>.</u>	
exclude files, 359–361	control flow nodes, 306-307	ingesting	
hadoop-env.sh file, 371	core-site.xml properties, 372-373	best practices, 77–78	

with Flume, 63-70 data storage and exchange MapR, 36-38 formats, 318-319 with RESTful interfaces, non-commercial Hadoop, 74-77 27-32 data structures in Pig Latin, 128 with Sqoop, 70-74 data transport encryption, 410 Hive, 169, 178-179 inspecting schema in Pig, 135 data warehouse offload use case, NameNode high availability, 7, 445-446-447 393-398 joining in Hive, 176-177 databases deployment modes, 18-20 loading conventional, 167 deprecation of property names, in Hive, 174 368 Hive, 169 in Pig, 131-133 dereferencing operator, 143 DataFrames, 234-236 multiple datasets (Pig) **DESCRIBE FUNCTION statement** DataFu. 158 cogrouping, 144-145 (Hive), 176 DataNodes, 13, 49-50 concatenating, 153 **DESCRIBE** statement block distribution, 359 cross joins, 154-155 in Hive, 172 block failure recovery, 48 joining, 145-151 in Pig, 135 properties for HDFS, 375-376 removing duplicates, 154 deserialization, 103 datatypes splitting, 155-156 dfsadmin command, 422-423 in Hadoop, 104 subtracting tuples, 154 dfs.blocksize property, 376 in Hive, 171, 187-190 nested FOREACH statements dfs.datanode.data.dir property, in Pig, 143-144 in Pig Latin, 129-130 375 ordering in Pig, 134 debugging. See troubleshooting dfs.datanode.du.reserved property, outputting in Hive, 180-181 decommissioning nodes from 375-376-376 clusters, 359-361 projecting and transforming in dfs.namenode.checkpoint.dir Pig. 134 deep storage, 444 property, 374-375 running queries in Pig, default configuration values, dfs.namenode.checkpoint.period 136-138 368-369 property, 374-375 data center, Hadoop integration deleting dfs.namenode.checkpoint.txns in, 443-444 directories in HDFS, 57 property, 374-375 data ingestion utilities, 301-303 files in HDFS, 57 dfs.namenode.edits.dir property, Flume, 302-303, See also deploying 373-374 Flume clusters dfs.namenode.name.dir property, Sqoop, 302. See also Sqoop Apache Ambari, 337-340 373-374 data locality, 2-3, 8, 229 in cloud, 38-41 dfs.replication property, 376 data masking in Hive, 196-199 Directed Acyclic Graphs (DAGs), Cloudera Manager, 33–34, data processing. See HBase; Hive; 340 215 MapReduce; Pig; Spark commercial Hadoop, directories in HDFS data science. See advanced 33-38.42 deleting, 57 analytics use case Hortonworks, 35-36

listing contents, 57 machine learning, 319-323 external tables, internal versus, paths, 56 NoSQL solutions, 305-306 extracting data with Regex SerDe, DISTINCT statement (Pig), 154 notebook platforms, 323-325 191 distributed computing, limitations search platforms, 306 of. 82 stream processing, 311-315 DistributedCache, 118-119 workflow tools, 311 document stores, 265 F edits file, 54 downloading Elastic Compute Cloud (EC2), files in HDFS, 57 38-39 failover, types of, 392-393 Oozie, 311 election analogy for MapReduce, failure recovery Drill, 208, 216-217 HDFS blocks, 48 EMR (Elastic MapReduce), 40 YARN, 283-284 drivers encryption MapReduce, 106-107, FairScheduler, 293-295 109-111 data at rest. 411 fault tolerance, 12 Spark, 226-227 data transport encryption, 410 MapReduce, 88 DROP PARTITION statement enterprise integration of Hadoop, master and slave nodes, 42 (Hive), 200 443-444. See also use cases with RDDs, 230-231 DStreams, 236-237, 312-313 environment variables for all YARN, 283-284 users, 59 DUMP statement (Pig), 138 feature engineering, 449-450 event lakes, 448 duplicates, removing in Pig, 154 federation in HDFS, 398-401 event processing use case, 7, fencing, 391-392 447-449 fields (Pig Latin), 128 exclude files, configuring, E FIFOScheduler, 293 359-361 file globs, 111 executing queries in Hive. EC2 (Elastic Compute Cloud), files in HDFS, 12, 46-48 203-204 deleting, 57 executors, Spark, 227 ecosystem projects, 4-5, 301. downloading, 57 EXPLAIN statement (Hive), 203 See also names of specific ingesting, 56 **EXPLODE function, 192** ecosystem projects reading and writing, extending MapReduce, 120 data ingestion utilities, 50-52, 60 extensions to Spark 301-303 FILTER statement (Pig), 133 GraphFrames, 239-240 data processing utilities. filtering data in Pig, 133, 150 303-304 GraphX, 239-240 Flink, 208 MLlib, 238-239, 450 data storage and exchange Flume, 63-70, 302-303 formats, 318-319 Spark Streaming, 236–237 architecture, 64-67 infrastructure and security, SparkR, 237-238 315-318 installing, 67-70

SparkSQL, 234-236

FOREACH statement (Pig), 134 Hadoop Distributed File System н (HDFS). See HDFS (Hadoop built-in functions, 136 Distributed File System) nesting, 143-144 H₂0, 321-322, 450 **Hadoop User Environment** fs.defaultFS property, 372 Hadoop (HUE). See HUE (Hadoop User fsimage file, 54 Cassandra with, 274-275 **Environment**) full outer joins, 145 cluster architecture, 17–18 hadoop-env.sh file, 371 fully-distributed mode, 18 commercial distributions, 5-6 hadoop-metrics2.properties file, functions core components, 3-4, 9 433-434 built-in functions in Pig, 136 deployment modes, 18-20 hadoop-metrics.properties file, text processing in Hive. ecosystem projects, 4-5 372 192-193-194 enterprise integration in hadoop.tmp.dir property, 377 UDFs. See UDFs (user-defined data center, 443-444 hardware requirements, 24-26 functions) explained, 2-3 hashing functions in Hive, future of NoSOL, 275 HBase with, 272 196-199 history of, 1-2 HAWQ, 208, 216 installing, 26-27 HBase, 265, 305 in cloud, 38-41 G architecture, 268-269 Cloudera Manager, 33-34 configuration, 383 GRANT statement (Hive), 170 commercial Hadoop, data model, 266 33-38, 42 graph processing with Spark, data processing in, 266-268 239-240 Hortonworks, 35-36 with Hadoop, 272 graph stores, 265 MapR, 36-38 **HUE HBase browser** GraphFrames, 239-240 non-commercial Hadoop, interface, 249 27 - 32GraphX, 239-240 installing, 269-272 MapReduce in, 95-96 GROUP ALL statement (Pig), 143 HDFS (Hadoop Distributed File requirements GROUP statement (Pig), 141-142 System), 3-4, 9 hardware, 24-26 ACLs, 53-54 grouping operating system, 23-24 data administration, 422-424 software, 26 in Hive. 176 archiving, 402 use cases, 6-8 in Pig, 141-142 blocks, 12, 46-48 data warehouse offload. multiple datasets in Pig. distribution of, 47 445-446-447 144-145 failure recovery, 48 event processing, grunt shell (Pig), 127-128 replication of, 47-48 447-449 caching, 401 predictive analytics, cluster processes, 11-14, 21

HUE Hive editor, 246-247 DataNodes, 49-50 on-disk structures, 54 NameNode, 48-49 permissions, 53-54, 406-407 InputFormats, 173-174 SecondaryNameNode, principles of, 45-46 metastore, 166-168 55.60 rack awareness, 387-389 OutputFormats, 173-174 configuration recovery, 54 query execution, 203-204 client properties, 376-377 replication, 46-48 query optimization common properties, safe mode, 54-55 bucketing, 201-202 372-373 snapshots, 401-402 partitioning, 199-201 DataNode properties, Trash directory settings, 57 SerDes. 173-174 375-376 WebHDFS, 74-76 tables, 169 NameNode properties, high availability analyzing data, 175 373-374 in HDFS, 55, 390-398 grouping and aggregating SecondaryNameNode data, 176 deploying, 393-398 properties, 374-375 internal versus external, failover types, 392-393 consistency, 54 173 fencing, 391-392 daemons, 21 joining data, 176-177 for YARN Resource directories loading data, 174 Manager, 280 deleting, 57 outputting data, 180-181 history listing contents, 57 text processing of big data, 1-2 paths, 56 data masking and hashing, of Hadoop, 1-2 federation, 398-401 196-199 Hive, 165-166, 303-304 files, 12, 46-48 functions for, 192-193 administration, 425-426 deleting, 57 regular expressions, authorization, 169-171 downloading, 57 190-192 built-in functions, 175-176 ingesting, 56 sentiment analysis, CLI, 168-169 reading and writing, 194-196 configuration, 381 50-52, 60 HiveQL, 165, 175 conventional databases high availability, 55, 390-398 HiveServer 2, 168 versus, 167 deploying, 393-398 Hortonworks, 6, 35-36 creating objects, 171–173 failover types, 392-393 HttpFS. 76-77 databases, 169 fencing, 391-392 HUE (Hadoop User Environment), datatypes HttpFS, 76-77 243-244, 258-260 complex, 187-190 interacting via HUE, 244-245 advantages of, 244 simple, 171 interfaces for, 56 authentication, 255 deploying and configuring, metadata, 12, 52-53, 60 authorization, 256 169, 178-179

Kafka, 313-315, 447

key value pairs, 83-84, 114

Kerberos, 411-414

configuring, 251–255	input/output types in Spark,	Java MapReduce API
HBase browser interface, 249	224-225	applications
HDFS file browser interface,	InputPaths, 111	compiling and packaging,
244–245	INSERT statement (Hive), 181	114–116
Hive editor interface, 246–247	inspecting schemas in Pig, 135	writing, 109–114
Impala query editor, 248	installing	Combiner functions, 117
installing, 251	Apache Ambari, 335, 351	DistributedCache, 118-119
logging in, 256-258	Cloudera Manager,	Hadoop datatypes, 104
Oozie editor, 250, 310	332–333, 351	InputFormats, 105
Pig editor, 247–248	Flume, 67–70	OutputFormats, 106
Solr interface, 249-250	Hadoop, 26–27	Partitioners, 117–118
Sqoop interface, 251	in cloud, 38–41	SequenceFiles, 105
YARN management, 245	Cloudera Manager, 33–34	serialization, 103
hue.ini file, 251-255	commercial Hadoop,	Java Virtual Machine (JVM), 19
	33–38, 42	JBOD (just a bunch of disks),
	Hortonworks, 35-36	25, 42
	MapR, 36-38	JDBC (Java Database
1	non-commercial Hadoop,	Connectivity), 71, 73
identifiers in Pig Latin 129 120	27–32	Job object, configuration and, 107
identifiers in Pig Latin, 128–129	HBase, 269-272	JOIN statement
ILLUSTRATE statement (Pig), 135	HUE (Hadoop User	in Hive, 176
immutability, 46	Environment), 251	in Pig, 149–150
Impala, 208, 211–214, 248	Pig, 126	joining
IMPORT statement (Pig), 161	Spark, 232-234	data in Hive, 176–177
include files, configuring,	internal tables, external	multiple datasets in Pig,
357-359	versus, 173	145–151
infrastructure projects, 315-318	IoT (Internet of Things), 448	joins, types of, 145
ingesting	IPython, 324	Jupyter, 324-325
data		JVM (Java Virtual Machine), 19
best practices, 77-78		,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,
with Flume, 63-70	J	
with RESTful interfaces,	-	17
74–77	Java classes, naming	K
with Sqoop, 70-74	conventions, 106	

conventions, 106

(JDBC), 71, 73

Java Database Connectivity

files in HDFS, 56

InputFormats, 105, 173-174

inner joins, 145

datatypes, 104
serialization, 103
key value stores, 265
keywords in Pig Latin, 129
killing
applications in YARN,
288–289
Hive queries, 203
Knox, 415–416

LATERAL VIEW clause, 193

left outer joins, 145

lazy evaluation in Spark, 231-232

L

LEFT SEMI JOIN statement (Hive), 177 LEVENSHTEIN function, 196 libraries in Pig, 157-158 linear scalability, 355-356 Lisp. 84-85 listing directory contents in HDFS, 57 listings accessing files in DistributedCache, 119 accessing HBase from PySpark, 272 accessing logs for completed applications, 430-431 accessing MRJobHistory Server via REST, 292

accessing SparkSQL

actions in Spark, 231

DataFrames, 235-236

ADD FILE and TRANSFORM statements, 187 ADD PARTITION statement, 200 adding files to DistributedCache from command line, 119 adding Hadoop environment variables for all users, 59 adjusting daemon log level, 429 adjusting log level programmatically, 430 Ambari blueprints, 339-340 Apache DataFu function library, 158 ARRAY datatype, 188 array() function, 188 assigning cluster ID when formatting NameNode, 399 assigning HDFS object permissions and ownership, 53 AvroMapper, 318-319 backing up NameNode's fsimage file, 424 balancer utility, 362 built-in functions in Pig, 136 changing Hive database contexts, 169 changing object permissions, 407 changing ownership of object in HDFS, 407 COGROUP statement, 145 comments in Pig, 131 common built-in functions included with Pig, 136

configuring CapacityScheduler, 296-297 configuring client side mount points for HDFS federation, 400 configuring connection to shared Hive metastore, 381 configuring FairScheduler, 295 configuring HA fencing, 392 configuring Hive authorization, 170 configuring SSL transport security for YARN, 287 CONTEXT NGRAMS function, 195 CREATE TABLE AS SELECT, 174 CREATE TABLE statement. 172 CREATE TABLE with bucketing, 201 CREATE TABLE with partitioning, 199 creating and updating Java MapReduce counters, 435 creating directory snapshot, 402 creating Hadoop archive, 402 creating keyspace and table in Cassandra using cqlsh, 274 creating roles in Hive, 170 creating snapshottable directory in HDFS, 402 creating table and inserting data in HBase, 267 CROSS statement, 155 custom Partitioner, 118 custom transform script written in Python, 186-187

declaring Combiner in Driver class, 117 declaring Partitioner in Driver class, 118 defining and calling Pig macro, 160-161 DESCRIBE FUNCTION statement, 176 DESCRIBE statement, 135, 172 dfs.blocksize configuration property, 376 dfs.datanode.data.dir configuration property, 375 dfs.datanode.du.reserved configuration property, 376 dfs.namenode.name.dir configuration property, 374 dfs.replication configuration property, 376 DISTINCT statement, 154 DROP PARTITION statement, 200 enabling Capacity Scheduler, 296 enabling FairScheduler, 294 enabling log aggregation, 379 enabling rack topology script, 389 enabling the capability to kill running jobs using UI, 288 encryption and decryption in Hive. 197 example log events, 428 EXPLAIN statement, 203 EXPLODE function, 192 file system ACLs, 406

FILTER statement, 133

final output (word count algorithm), 95 <final> tag, 370 Flume agent channel configuration, 67 Flume agent sink configuration, 66 Flume agent source configuration, 66 FOREACH statement, 134 fs.defaultFS setting, 372 GRANT statement, 170 GROUP ALL statement, 143 GROUP statement, 142 grouping and aggregating data in Hive, 176 Hadoop configuration needed to join cluster, 357 hadoop-metrics2.properties file, 433-434 hadoop.tmp.dir configuration property, 377 hbase-site.xml configuration file. 383 Hive built-in functions, 175 hue.ini file, 252-254 ILLUSTRATE statement, 135 IMPORT statement, 161 importing all tables from database using Sqoop, 73 input to map task, 93 INSERT statement, 181 installing ambari-server package using EPEL, 335 intermediate data sent to Reducer, 94 item based recommender in Mahout, 320

JOIN statement, 149-150, 176 killing Hive query, 203 killing YARN application using yarn command, 290 LATERAL VIEW clause, 193 leaving safemode, 423 LEFT SEMI JOIN statement, 177 LEVENSHTEIN function, 196 LOAD statement, 174 local execution in Hive, 204 log4j.properties file, 371, 427 manual failover, 392 MAP datatype, 190 mapreduce.framework.name configuration property, 380 mapreduce.jobhistory.address configuration property, 380 MASK function, 197 MD5 function, 197 MSCK REPAIR TABLE command, 201 multiple Pig parameters in text file, 160 nested FOREACH statements. 144 NGRAMS function, 195 ORDER statement, 134 OVERWRITE option, 174 PageRank algorithm in Spark, 240 parameterized HiveQL query, 185 Parquet storage, 210 PARSE_URL function, 198

PARSE_URL_TUPLE function, 199
passing parameters to Pig script, 160
passing Pig parameters using text file, 160
passing shell command return values to Pig script, 160
performing snapshot dff, 402
Pig dryrun option, 161
Pig identifiers, 129
Pig program, 126
Pig program variables sourced from parameters, 159
Pig properties file, 382
Pig reserved keywords, 129
Pig UDF written in Jython, 157
PiggyBank function library, 158
PigStorage LOAD function, 132
POSEXPLODE function, 193
rack topology script, 389
RDDs in Spark, 228
reading MapReduce counters, 435
referencing Hive objects, 169
refreshing CapacityScheduler
configuration, 297
Regex SerDe, 191
REGEXP_EXTRACT
function, 191
REGEXP_REPLACE
function, 192
registering and using Pig UDF, 157

REST access to HDFS using WebHDFS, 74
RESTful HDFS access using HttpFS, 77
REVOKE statement, 170
running Hive queries in batch mode, 169
sample datasets, 142
sample Oozie workflow document, 307–308
scanning HBase table, 267
schematool usage to initialize metastore schema, 425
securing NameNode's web UI using HTTPS, 408
SELECT statement, 175
selecting ARRAY elements, 188
selecting MAP values, 190
selecting STRUCT elements, 189
SENTENCES function, 194
SET statement, 186
setting Hive variable via the command line, 186
setting Spark configuration properties programmatically, 382
SHOW PARTITIONS statement, 200
SOUNDEX function, 196
Spark configuration properties, 382
Spark MLlib, 320–321
specifying log4j.properties file for application, 428

tasks from command line, 110 SPLIT function, 192 SPLIT statement, 155-156 Sgoop tools, 72-73 Sgoop2 shell, 302 starting services in Hadoop, 422 STREAM statement, 159 STRUCT datatype, 189 submitting Oozie workflow using command line client, 309 submitting Spark application, 224 submitting streaming MapReduce job, 122 SUBTRACT statement, 154 SUM function, 143 supplying schemas for bags in Pig, 132-133 TABLESAMPLE operator, 202 Timeline Server REST API, 293 ToolRunner command line generic options, 110 training a decision tree model with Spark MLlib, 239 transformations in Spark, 231 transport security for intermediate data, 410 UNION statement, 153 updating cell in HBase table, 268 updating data in Cassandra table using Spark, 275

specifying number of Reduce

valid and invalid Pig log aggregation in YARN, mapred-site.xml file, 379 identifiers, 129 290-291, 379 MapReduce. See also Java ViewFS, 400 log files MapReduce API viewing YARN logs using yarn within applications, 430-431 applications command, 290-291 for daemons, 428-430 compiling and packaging, web log schema using RegEx 114-116 Log4i, 426-427 SerDe, 198 components of, 106-108 log4j.properties file, 371, word count mapper in 427-428 writing in Java, 109-114 Python, 121 Log4j, 426-427 asymmetry, 89-90 word count reducer in log4j.properties file, 371, Combiner functions, Pvthon, 121 427-428 88-89, 117 WordCount Driver class, 109 logging in to HUE, 256-258 counters, 434-439 WordCount Mapper class, 111 Lucene, 306 design goals of, 83 WordCount Reducer LVM (Logical Volume election analogy, 91-92 class, 113 Manager), 24 EMR. 40 yarn command, 289 extending, 120 yarn command usage to confault tolerance, 88 trol YARN applications, 425 Google whitepapers on, 81 M YARN container related conin Hadoop, 95-96 figuration parameters, 282 machine learning, 319-323 history with YARN, 291-293 yarn.app.mapreduce. with H₂0, 321-322 kev value pairs, 83-84 am.staging-dir configuration property, 380 with Mahout, 320 LocalJobRunner mode, 19-20 yarn.nodemanager.auxpredictive analytics use case, Map phase, 85-86, 98 services configuration 448-450 Map-only applications, 90-91 property, 378 Presto, 323 motivation for, 82 yarn.nodemanager.local-dirs with Spark, 238-239, Partitioners, 86-87, 117-118 configuration property, 378 320-321 programming model, 84-85 yarn.nodemanager.log-dirs macros in Pig, 160-161 properties for YARN, 379-380 configuration property, 379 Mahout, 320, 450 Reduce phase, 87-88, 98-99 LOAD statement (Hive), 174 MAP datatype, 189-190 Shuffle and Sort phase, loading data Map phase of MapReduce, 87.98 in Hive. 174 85-86, 98 speculative execution, 89-90 in Pig, 131-133 Map-only MapReduce applications, Sgoop and, 71-72 local mode, running Hive, 90-91 word count algorithm, 92-95 203-204 Mappers (in MapReduce), MapReduce Streaming API, LocalJobRunner mode, 19-20 107-108, 111-113 120-122 MapR, 6, 36-38

mapreduce.framework.name property, 380 mapreduce.jobhistory.address property, 380 MASK function, 197 master nodes

fault tolerance, 42
hardware requirements, 24
master-slave cluster
architecture, 12
mathematical operators in Pig
Latin, 130-131
MBeans, 432-434
metadata in HDFS, 12, 52-53, 60
metastore (Hive), 166-168,
425-426
metrics framework, 432-434
MLIib, 238-239, 320-321, 450

Apache Ambari, 347–350 Cloudera Manager, 347–350 counters, 434–439 MBeans, 432–434

MRJobHistory Server, 291–292
MSCK REPAIR TABLE command
(Hive), 201

multiple datasets

monitoring clusters

cogrouping in Pig, 144–145 concatenating in Pig, 153 cross joins in Pig, 154–155 joining in Pig, 145–151 removing duplicates in Pig, 154 splitting in Pig, 155–156

subtracting tuples in Pig, 154

N

name value pairs. See key value pairs

NameNodes, 12-13, 48-49

backing up, 423
federation, 398–401
high availability, 390–398
deploying, 393–398
failover types, 392–393
fencing, 391–392
metadata, 52–53, 60
properties for HDFS, 373–374
safe mode, 54–55
as single point of failure, 55

nameservices, 394 naming conventions

bags (Pig Latin), 133 Java classes, 106 in Pig Latin, 128–129, 131

nested FOREACH statements (Pig), 143–144

networks, hardware requirements, 26 n-grams, 156–157, 194–195 NGRAMS function, 194–195 NodeManager, 16, 96, 281 failure recovery, 283 properties for YARN, 377–379

nodes, 4

action nodes, 307 adding to cluster, 356–359 control flow nodes, 306–307 decommissioning from cluster, 359–361 master nodes

fault tolerance, 42 hardware requirements, 24 slave nodes

co-locating, 21
fault tolerance, 42
hardware requirements, 25
non-commercial Hadoop installing,

normalizing text, 195 NoSQL, 305–306. See also Cassandra; HBase

27-32

Bigtable and, 263–264 characteristics of, 264–265 future of, 275 implementations of, 275 types of systems, 265 notebook platforms, 323–325



object identifiers in Pig Latin, 128–129 objects (Hive), creating, 171–173 ODPi (Open Data Platform initiative), 6 on-disk structures in HDFS, 54 on-platform analytics, 450 Oozie

architecture, 309 downloading, 311 HUE Oozie editor, 250, 310 workflows, 306–308, 309–310

Open Data Platform initiative (ODPi), 6

projecting data in Pig, 134

467

operating system requirements, perimeter security, 414-416 STREAM statement, 158-159 23-24 permissions in HDFS, 53-54, **UDFs** 406-407 operators community function mathematical operators in Pig persistence with RDDs, 229 libraries, 157-158 Latin. 130-131 persistent staging offload, registering, 157 relational operators in Pig 445-446 writing, 156-157 Latin, 130 Pig. 125-126, 304 WordCount example, 151 optimizing applications Pig Latin, 125 joins in Pig, 147 parameterizing, 159-160 comments, 131 queries in Hive, 199-202 reusing code, 160-161 data structures, 128 ORC (Optimized Row Columnar) built-in functions, 136 mathematical operators, format for columnar storage, configuration, 381-382 130-131 210-211 data object identifiers, 128-129 ORDER statement (Pig), 134 aggregating, 143 program flow, 128 ordering data in Pig. 134 filtering, 133, 150 relational operators, 130 outer joins, 145 grouping, 141–142 simple datatypes, 129–130 OutputFormats, 106, 173-174 inspecting schema, 135 statements, 131 outputting data in Hive, 180-181 loading, 131-133 PiggyBank, 157-158 nested FOREACH state-Pivotal HAWQ, 208, 216 ments, 143-144 planning applications in Spark, P ordering, 134 projecting and policies, access control, 416-418 packaging MapReduce transforming, 134 POSEXPLODE function, 193 applications, 114-116 running queries, 136-138 precedence rules in configuration, PageRank, 240 grunt shell, 127-128 369-370 parallel execution in Hive, 203 HUE Pig editor, 247-248 predictive analytics use case, 7-8, 448-450 parameterizing installing, 126 applications in Pig, 159-160 preemption, 295 modes, 127 queries in Hive, 185-186 Presto, 208, 323 multiple datasets pricing AWS, 40-41 Parquet format for columnar cogrouping, 144-145 storage, 209-210 processing data. See HBase; Hive; concatenating, 153 parsing URLs in Hive, 198-199 MapReduce; Pig; Spark cross joins, 154-155 Partitioners, 86-87, 117-118 programming interfaces to Spark, joining, 145-151 222-225 partitioning in Hive, 199-201 removing duplicates, 154 programming model of paths for directories, 56 splitting, 155-156 MapReduce, 84-85 performance benchmarking subtracting tuples, 154

Hadoop, 362-365

properties

client properties for HDFS, 376-377 common properties for HDFS, 372-373 for YARN, 377 DataNode properties for HDFS, 375-376 MapReduce properties for YARN, 379-380 NameNode properties for HDFS, 373-374 NodeManager properties for YARN, 377-379 SecondaryNameNode properties for HDFS, 374-375 property names, deprecation, 368 pseudo-distributed mode, 19 PySpark, 223

Q

queries

custom scripts in Hive, 186–187 executing in Hive, 203–204 optimizing in Hive bucketing, 201–202 partitioning, 199–201 parameterizing in Hive, 185–186 running in Pig, 136–138

R

rack awareness in HDFS.

387-389 Ranger, 417 RBAC (role-based access control), 416-418 RDDs (Resilient Distributed Datasets), 227-229 actions, 231 data locality, 229 fault tolerance, 230-231 lineage, 229-230 persistence and re-use, 229 transformations, 231 reading files in HDFS, 50-52 rebalancing clusters, 361-362 RecordReader objects, InputFormats, 105 records in MapReduce, 83-84 recovery in HDFS, 54 Reduce phase of MapReduce, 87-88, 98-99 Reducers (in MapReduce), 108, 113-114 Regex SerDe, 191 REGEXP EXTRACT function, 191 REGEXP_REPLACE function, 192 registering UDFs in Pig, 157 regular expressions in Hive, 190-192 relational operators in Pig Latin. 130 relations (Pig Latin), 128 removing

duplicates in Pig, 154

tuples in Pig, 154

replication in HDFS, 46-48 requirements

hardware, 24–26 operating system, 23–24 software. 26

Resilient Distributed Datasets (RDDs). See RDDs (Resilient Distributed Datasets) ResourceManager, 15, 96,

280, 284

ResourceManager UI, 286–289

ResourceRequests, 281–283

resources for information, 440

RESTful interfaces, data ingestion

with, 74–77
re-use with RDDs, 229
reusing code in Pig, 160–161
REVOKE statement (Hive), 170
right outer joins, 145
role-based access control (RBAC),
416–418

running

applications on YARN, 96–99, 281–284 Hive in local mode, 203–204 queries in Pig, 136–138

S

S3 (Simple Storage Service), 39
safe mode in HDFS, 54–55
sbin directory scripts, 422
Scala, 224
scaling Hadoop
adding nodes to cluster,

benchmarking, 362-365 Sentry, 418 software requirements, 26 decommissioning nodes from Solr, 249-250, 306 SequenceFiles, 105 cluster, 359-361 SerDes (Serializer/Deserializer), SOUNDEX function, 196 linear scalability, 355-356 173-174. 191 sources (Flume agents), 65-66 rebalancing clusters, 361-362 serialization, 103 Spark, 279-280, 304 when to scale, 356 service management accumulators, 436 scheduling applications Apache Ambari, 344-345 application support, 222 in Spark, 226 Cloudera Manager, 340-342 architecture, 225 in YARN, 293-298 services drivers, 226-227 schema-on-read systems, 3 starting, 421-422 executors, 227 schema-on-write systems, 3 stopping, 421-422 configuration, 382 schemas, inspecting in Pig, 135 set operations (Pig) extensions scripts, custom scripts in Hive, CROSS statement, 154-155 GraphFrames, 239-240 186-187 DISTINCT statement, 154 GraphX, 239-240 search platforms, 306 SPLIT statement, 155-156 MLlib. 238-239. SecondaryNameNode, 13, 55, 60, SUBTRACT statement, 154 320-321, 450 374-375 UNION statement, 153 Spark Streaming, 236-237 security SET statement (Hive), 186 SparkR, 237-238 Apache Ambari, 351 severity levels (Log4j), 426 SparkSQL, 234-236 authentication, 411 shared nothing, 3, 9 installing, 232-234 authorization, 411 **SHOW PARTITIONS statement** lazy evaluation, 231-232 Cloudera Manager, 351 (Hive), 200 programming interfaces, ecosystem projects, 315-318 shuffle, 232 222-225 encryption Shuffle and Sort phase of RDDs. 227-229 data at rest, 411 MapReduce, 87, 98 actions, 231 data transport encryption, simple datatypes. See datatypes data locality, 229 410 Simple Storage Service (S3), 39 fault tolerance, 230-231 HDFS permissions, 406-407 single point of failure (SPOF), 55 lineage, 229-230 Kerberos, 411-414 sinks (Flume agents), 66-67 persistence and re-use, perimeter security, 414-416 229 slave nodes RBAC (role-based access co-locating, 21 transformations, 231 control), 416-418 fault tolerance, 42 Spark Streaming, 236-237, for web Uls. 407-409 312-313 hardware requirements, 25 SELECT statement (Hive), 175 SparkContext, 226 snapshots in HDFS, 401-402 SENTENCES function, 194 SparkR, 237-238 software distribution with sentiment analysis in Hive, Cloudera Manager, 332 SparkSQL, 234-236

STRUCT datatype, 189

spark-submit command, 224 SUBTRACT statement (Pig), 154 transforming speculative execution of SUM function (Pig), 143 data in Pig, 134 MapReduce, 89-90 symbolic notation for RDDs. 231 SPLIT function, 192 permissions, 406 Transparent Encryption, 411 SPLIT statement (Pig), 155-156 Trash directory settings, 57 splitting multiple datasets in Pig, troubleshooting 155-156 best practices for, 440 Т SPOF (single point of failure), 55 clusters SQL on Hadoop, 304 tables (Hive), 169 Apache Ambari, 347-350 Apache Drill, 216-217 Cloudera Manager, analyzing data, 175 Apache HAWQ, 216 347-350 grouping and aggregating explained, 207-208 data, 176 log files for Impala, 211-214 internal versus external, 173 within applications, Tez. 214-216 430-431 joining data, 176-177 Sgoop, 70-74, 302 daemon logs, 428-430 loading data, 174 custom connectors, 73 Log4i, 426-427 outputting data, 180-181 log4j.properties file. HUE Sqoop interface, 251 TABLESAMPLE operator 427-428 MapReduce and, 71-72 (Hive), 202 resources for information, 440 Sqoop2 (Sqoop-as-a-Service), task failure, YARN, 283 73-74, 302 with web UIs, 431-432 Terasort, 363-365 tools, 72-73 tuples (Pig Latin), 128, 154 text processing in Hive Sqoop2 (Sqoop-as-a-Service), data masking and hashing, 73-74, 302 196-199 SSL security for YARN, 287 functions for, 192-193-194 U StandbyNameNode, 13 regular expressions, 190-192 starting services, 421-422 sentiment analysis, 194-196 UDFs (user-defined functions) statements in Pig Latin, 131 TextInputFormat, 93, 94 community function libraries in stopping services, 421-422 Tez, 208, 214-216 Pig, 157-158 STORE command (Pig), 138 Thrift, 319 registering in Pig, 157 Storm, 311-312 tiered storage, 444 writing in Pig, 156-157 stream processing, 311-315 Timeline Server, 292-293 UNION statement (Pig), 153 Kafka. 313-315 tokenization, 93 uploading. See ingesting Spark Streaming, 312–313 tools in Sqoop, 72-73 URL parsing in Hive, 198-199 Storm, 311-312 **TPC (Transaction Processing** use cases, 6-8 STREAM statement (Pig), Performance Council), 363 data warehouse offload, 158-159 TRANSFORM operator (Hive), 445-446-447

event processing, 447–449 predictive analytics, 448–450 user-defined functions. See UDFs (user-defined functions)

V

ViewFS, 400



web log parsing in Hive, 198-199 web UIs

security for, 407–409 troubleshooting with, 431–432

WebHDFS, 74–76 word count algorithm, 92–95 workflows

in Oozie, 306–308, 309–310 tools for, 311 Writable interface, 104

WritableComparable interface, 104

writing

files in HDFS, 50–52, 60

MapReduce applications,
109–114

Driver code, 109–111

Mapper code, 111–113

Reducer code, 113–114

UDFs in Pig, 156–157



YARN (Yet Another Resource Negotiator), 3-4, 9

administration, 424–425
application scheduling,
293–298
CLI usage, 289–290
cluster processes, 14–17, 21,
279–281
configuration, 285–286
common properties, 377
MapReduce properties,
379–380
NodeManager properties,

daemons, 21 job management via HUE, 245

377-379

log aggregation, 290–291, 379
MapReduce history, 291–293
ResourceManager UI, 286–289
running applications on, 96–99, 281–284

varn application commands, 424

yarn rmadmin commands, 424
yarn.app.mapreduce.am.stagingdir property, 380
yarn.nodemanager.aux-services
property, 378
yarn.nodemanager.local-dirs property, 377–378
yarn.nodemanager.log-dirs property, 378–379
yarn.resourcemanager.hostname
property, 377
yarn-site.xml properties, 377



Zeppelin, 323-324 ZooKeeper, 315-318