



Data Science With Python

DATA AND ARTIFICIAL INTELLIGENCE



Python Integration with Hadoop MapReduce and Spark

Learning Objectives

By the end of this lesson, you will be able to:

- Explain why Python should be integrated with Hadoop
- Outline the ecosystem and architecture of Hadoop
- Explain the functioning of MapReduce
- Discuss Apache Spark functions and their benefits
- Write Python programs for Hadoop operations



Quick Recap: Need for Real-Time Analytics

We have seen how big data is generated and understood that to extract insights, proper analysis of data is more important than its size.



Quick Recap: Need for Real-Time Analytics

Real-time analytics is the rage right now because it helps extract information from different data sources almost instantly.

Date	Description	Deposit	Withdrawal	Balance	_]
Apr 1	ATM Post Debit		100	\$200,000	
Apr 2	Paypal Transfer 231054	200		\$202,000	
Apr 3	Simplilearn course fee		150	\$200,500	
Apr 4	Starluck Café		210	\$198,400	
Apr 5	Walcart TX		230	\$196,100	
Apr 6	ebuy swiss watch 239		250	\$193,600	
Apr 7	Caterpallor black boots men		270	\$190,900	
Apr 8	Halo blue shirt 831		160	\$189,300	
					-

Information source; overall patterns not clearly visible



Quick Recap: Need for Real-Time Analytics













Demographics

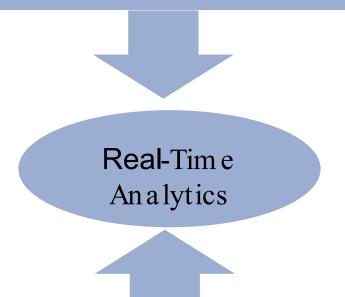
Orders

Products

Visits/Channels

Social Media Sentiments

Customer Support and Surveys





Higher Conversion



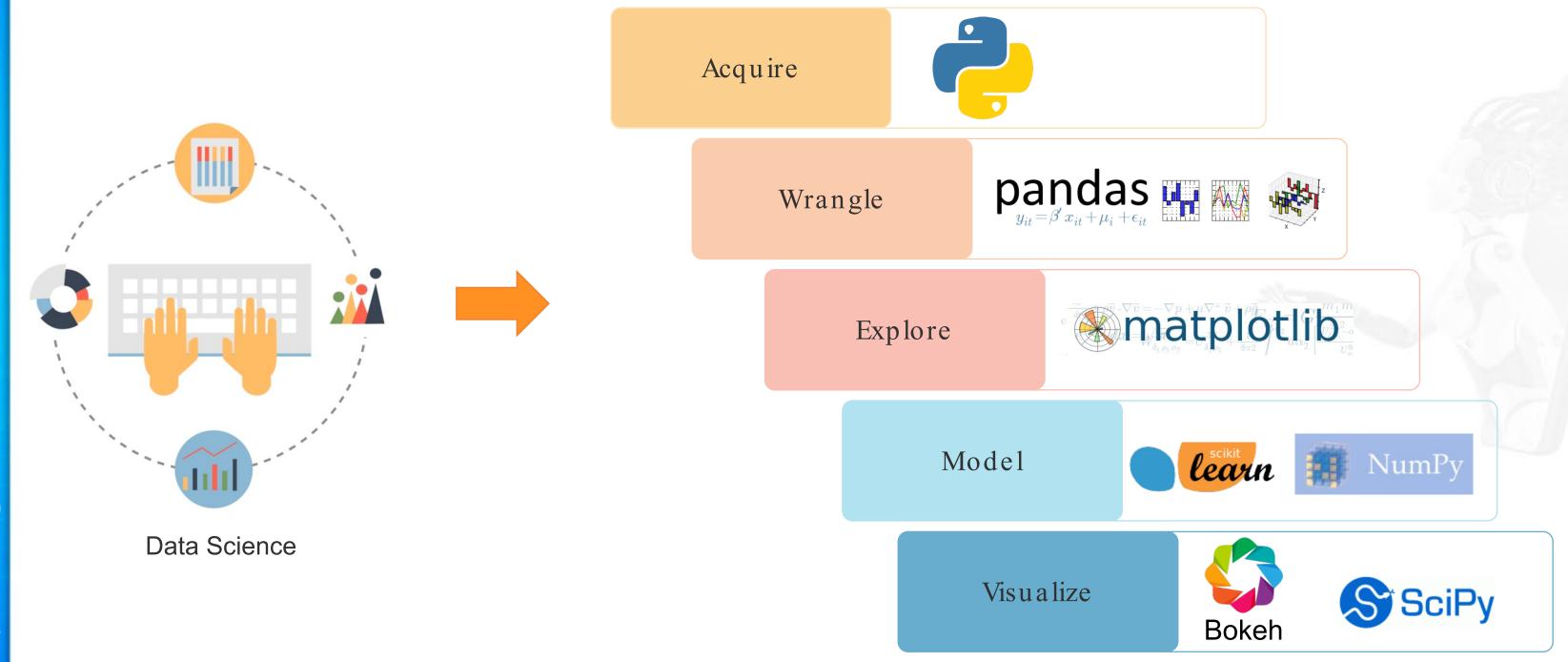




Increased Sales

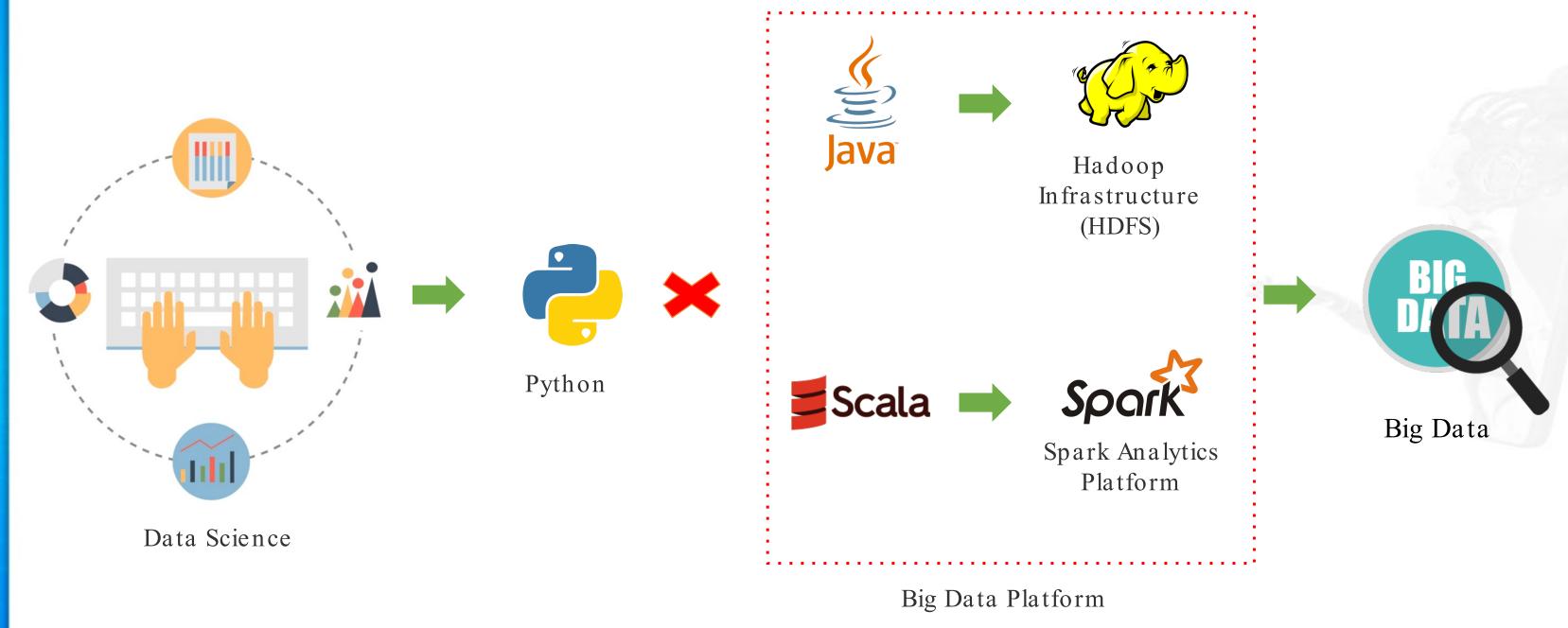
Quick Recap: Why Python

Data Scientists all over the world prefer Python because it is an easy -to-use language that has diverse libraries required for analytics.



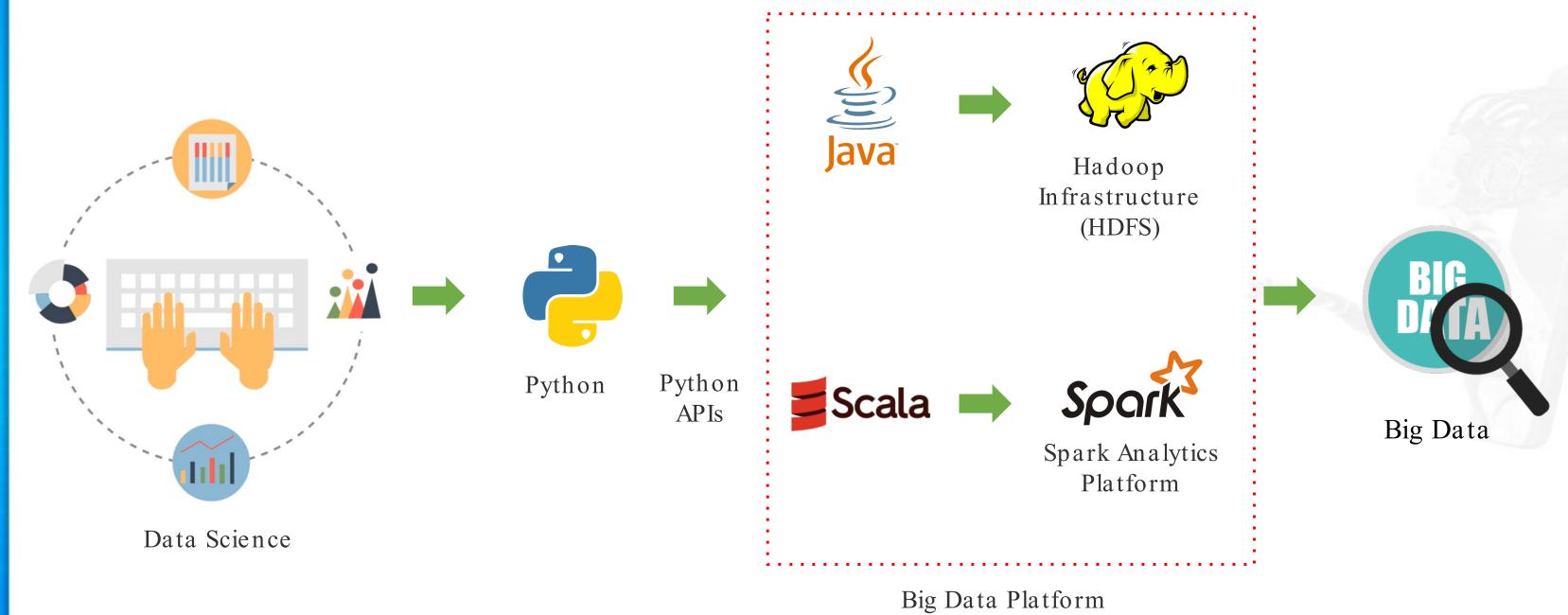
Disparity in Programming Languages

However, Big Data can only be accessed through Hadoop which is completely developed and implemented in Java. Also, analytics platforms are coded in different programming languages.



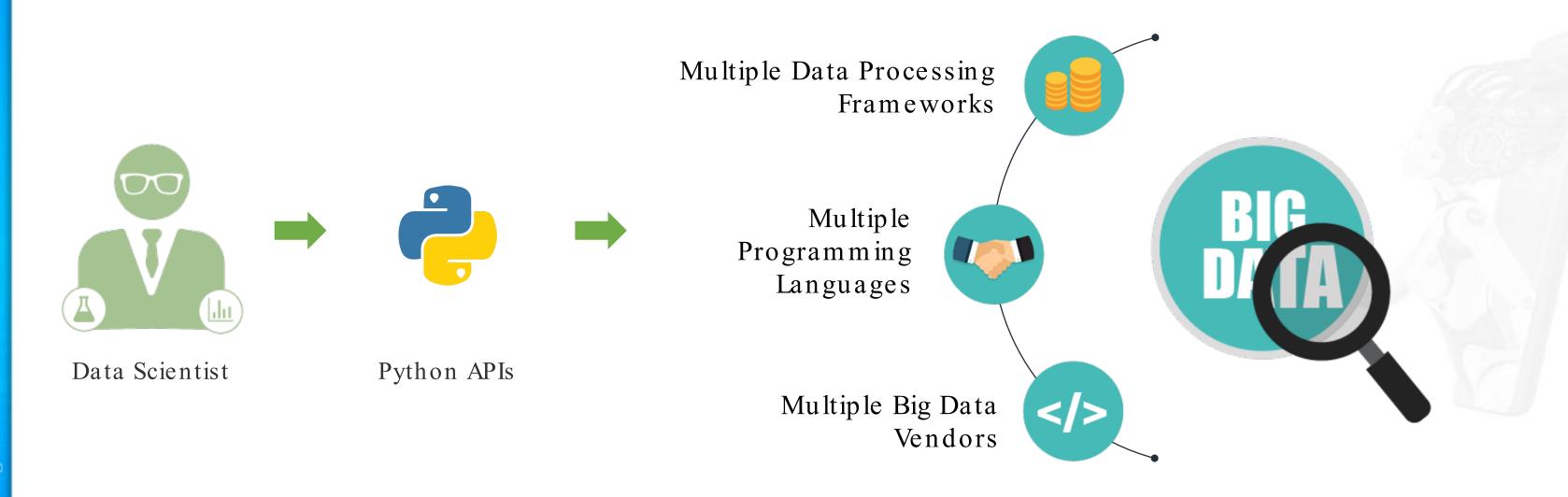
Integrating Python with Hadoop

As Python is a Data Scientist's first language of choice, both Hadoop and Spark provide Python APIs that allow easy access to the Big Data platform.



Need for Big Data Solutions in Python

There are several reasons for creating Big Data solutions in Python.



Hadoop: Core Components

Hadoop





HDFS

(Hadoop Distributed File System)

- It is responsible for storing data on a cluster
- Data is split into blocks and distributed across multiple nodes in a cluster
- Each block is replicated multiple times
 - o Default is 3 times
 - o Replicas are stored on different nodes

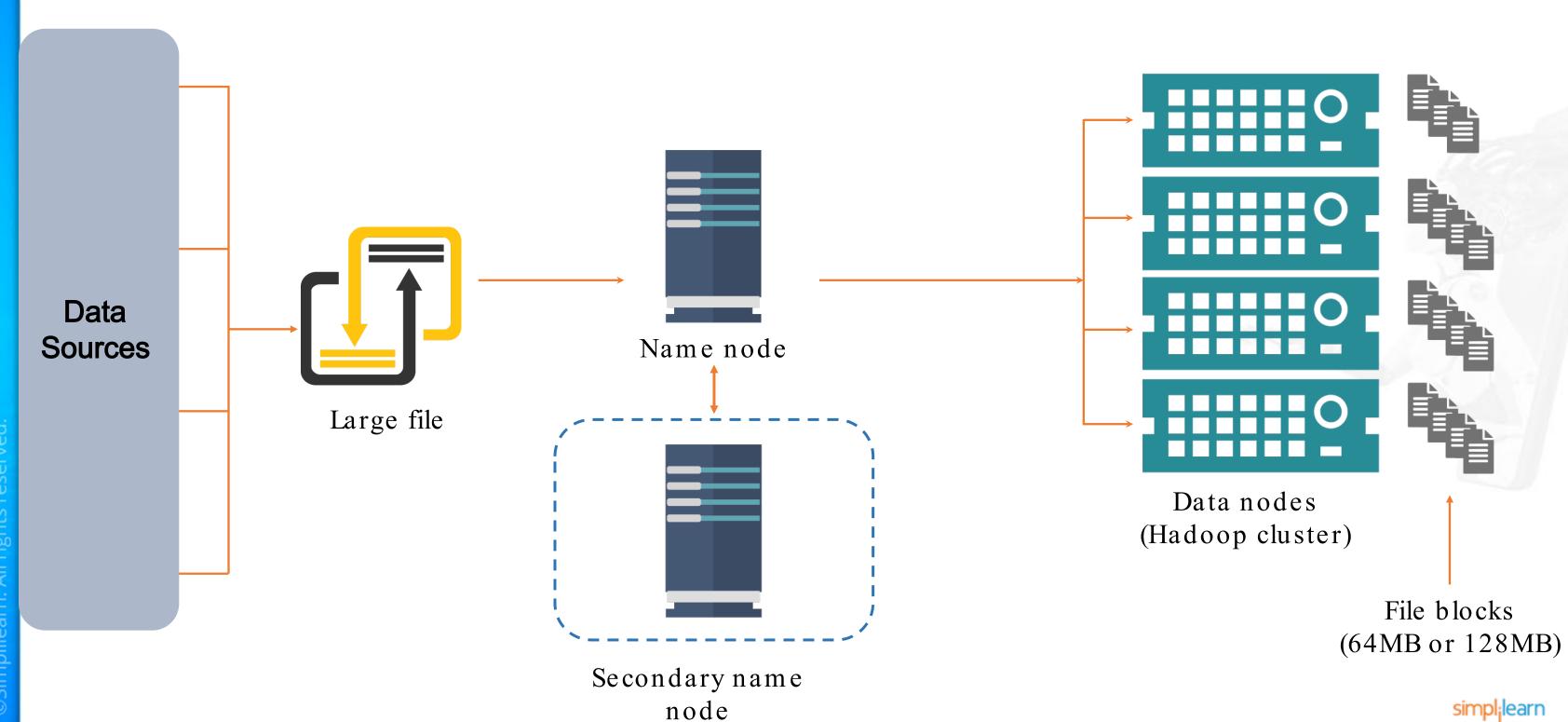


- MapReduce is a data processing framework to process data on the cluster
- Two consecutive phases: Map and Reduce
- Each map task operates on discrete portions of data
- After map, reduce works on the intermediate data distributed on nodes



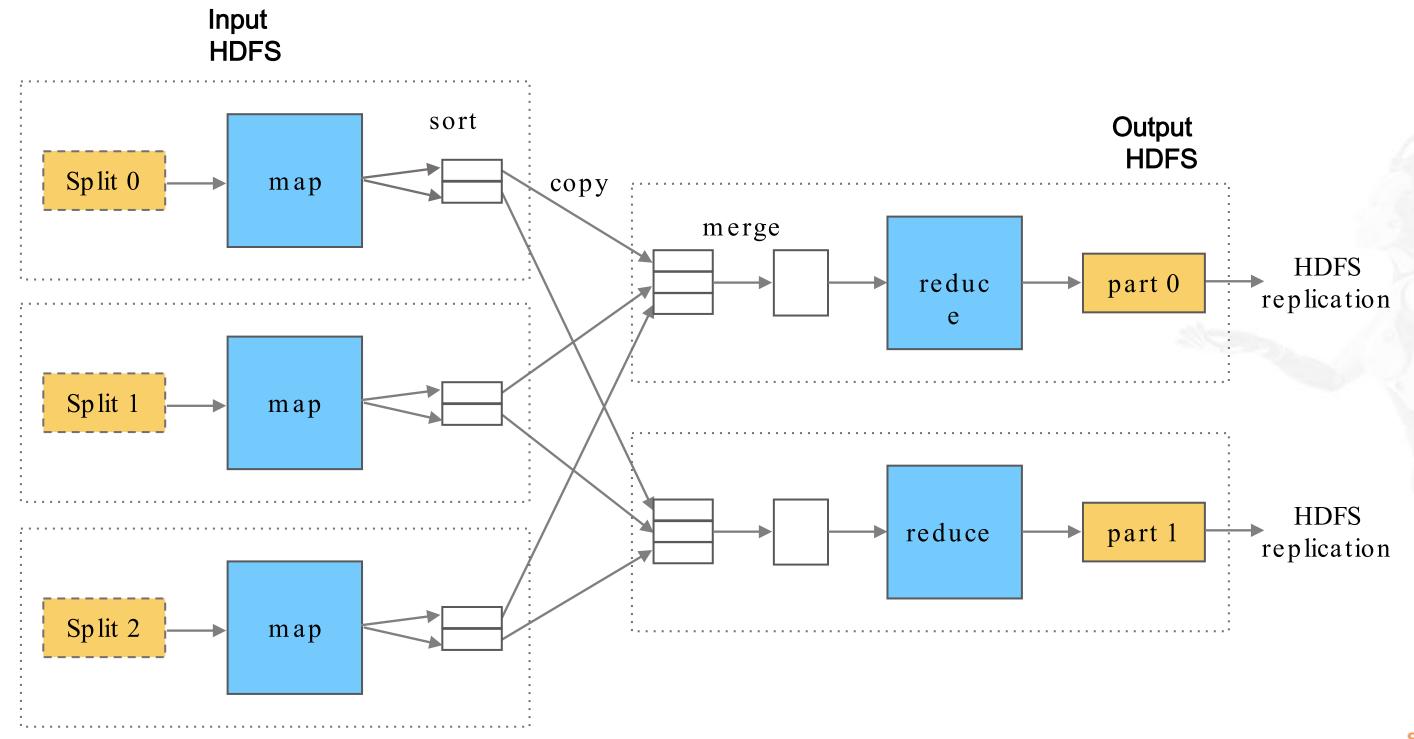
Hadoop: System Architecture

This example illustrates the Hadoop system architecture and the ways to store data in a cluster.



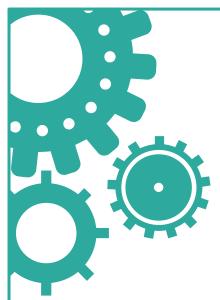
MapReduce

The second core component of Hadoop is MapReduce, the primary framework of the HDFS architecture.



MapReduce: Mapper and Reducer

Let us discuss the MapReduce functions, mapper and reducer, in detail.



Mapper

- Mappers run locally on the data nodes to avoid the network traffic.
- Multiple mappers run in parallel processing a portion of the input data.
- The mapper reads data in the form of key-value pairs.
- If the mapper write generates an output, it is written in the form of key -value pairs.

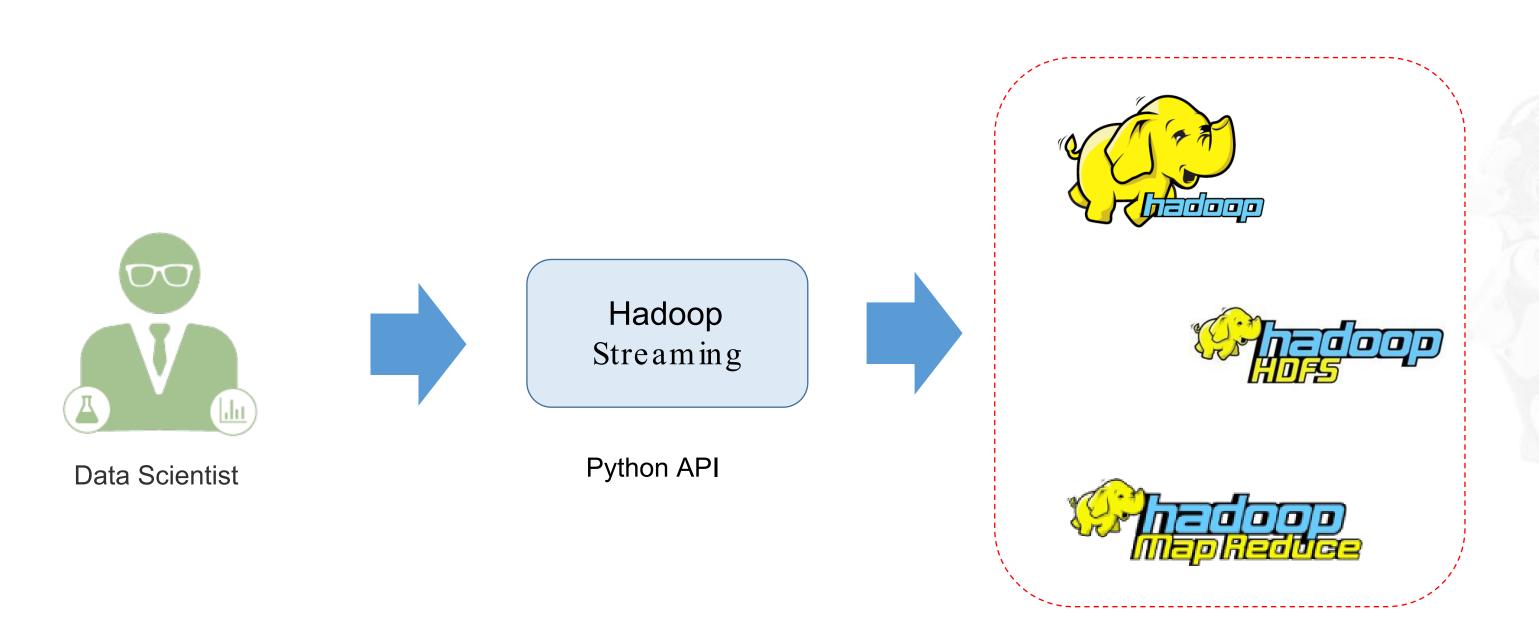
Reducer

- All intermediate values for a given intermediate key are combined together into a list and given to a reducer.
- This step is known as **shuffle** and **sort**.
- The reducer outputs either zero or more final key-value pairs. These are written to HDFS.



Hadoop Streaming: Python API for Hadoop

Hadoop Streaming acts like a bridge between your Python code and the Java -based HDFS, and lets you seamlessly access Hadoop clusters and execute MapReduce tasks.



Mapper in Python

Python supports map and reduce operations:

Suppose you have list of numbers you want to square = [1, 2, 3, 4, 5, 6]

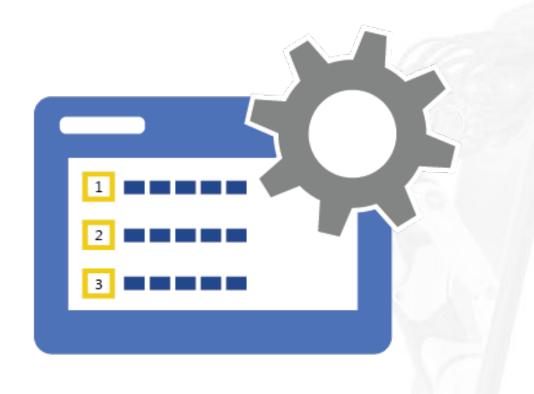
Square function is written as follows:

def square(num):

return num * num

You can square this list using the following code: squared_nums = map(square, <u>numbers</u>)

Output would be: [1, 4, 9, 16, 25, 36]



Reducer in Python

Suppose you want to sum the squared numbers:

[1, 4, 9, 16, 25, 36]

Use the **sum** function to add two numbers def sum (a, b):
return a + b

You can now sum the numbers using the **reduce** function

Import functools as f

sum_squared = f. reduce(sum, a)

Output would be:

[91]

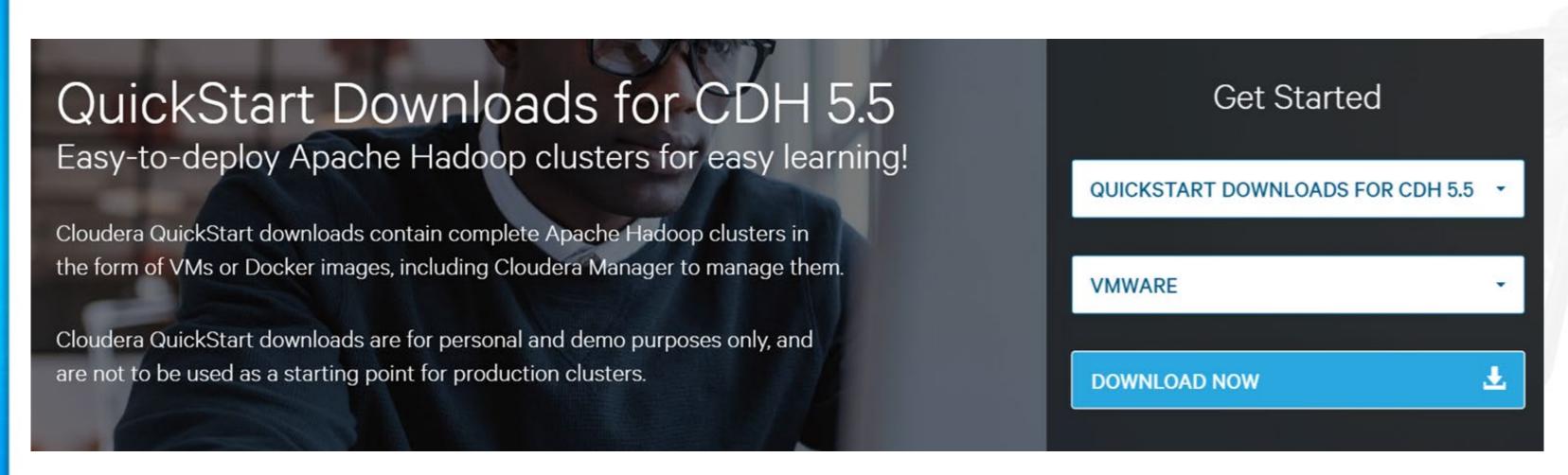


Setting Up Cloudera Quick Start Virtual Machine

Cloudera provides enterprise-ready Hadoop Big Data platform which supports Python as well.

To set up the Cloudera Hadoop environment, visit the Cloudera link:

http://www.cloudera.com/downloads/quickstart vms/5 -7.htm1



Cloudera recommends that you use 7-Zip to extract these files. To download and install it, visit the link: http://www.7-zip.org/

Cloudera Quick Start VM: Prerequisites

- These 64-bit VMs require a 64 -bit host OS and a virtualization product that can support a 64 -bit guest OS.
- To use a VMware VM, you must use a player compatible with WorkStation 8.x or higher:
 - Player 4.x or higher
 - Fusion 4.x or higher
- Older versions of WorkStation can be used to create a new VM using the same virtual disk (VMDK file), but some features in VMware tools are not available.
- The amount of RAM required varies by the run -time option you choose

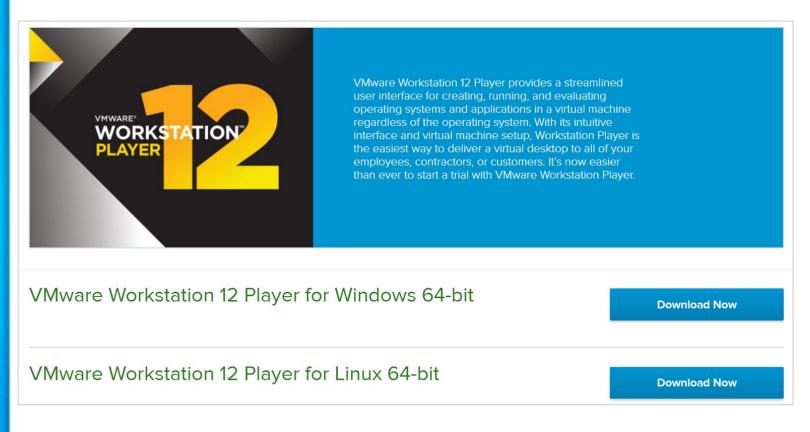
CDH and Cloudera Manager Version	RAM Required by VM		
CDH 5 (default)	4+ GiB*		
Cloudera Express	8+ GiB*		
Cloudera Enterprise (trial)	10+ GiB*		



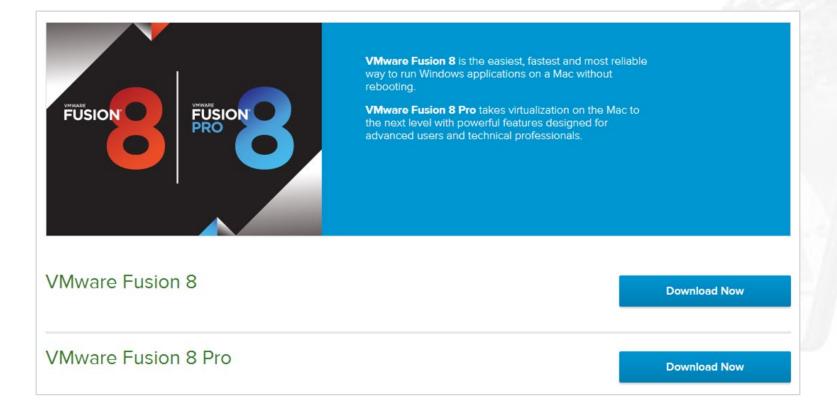
Launching VMware Image

To launch the VMware, visit the VMware link:

https://www.vmware.com/products/player/pla yerpro-evaluation.html



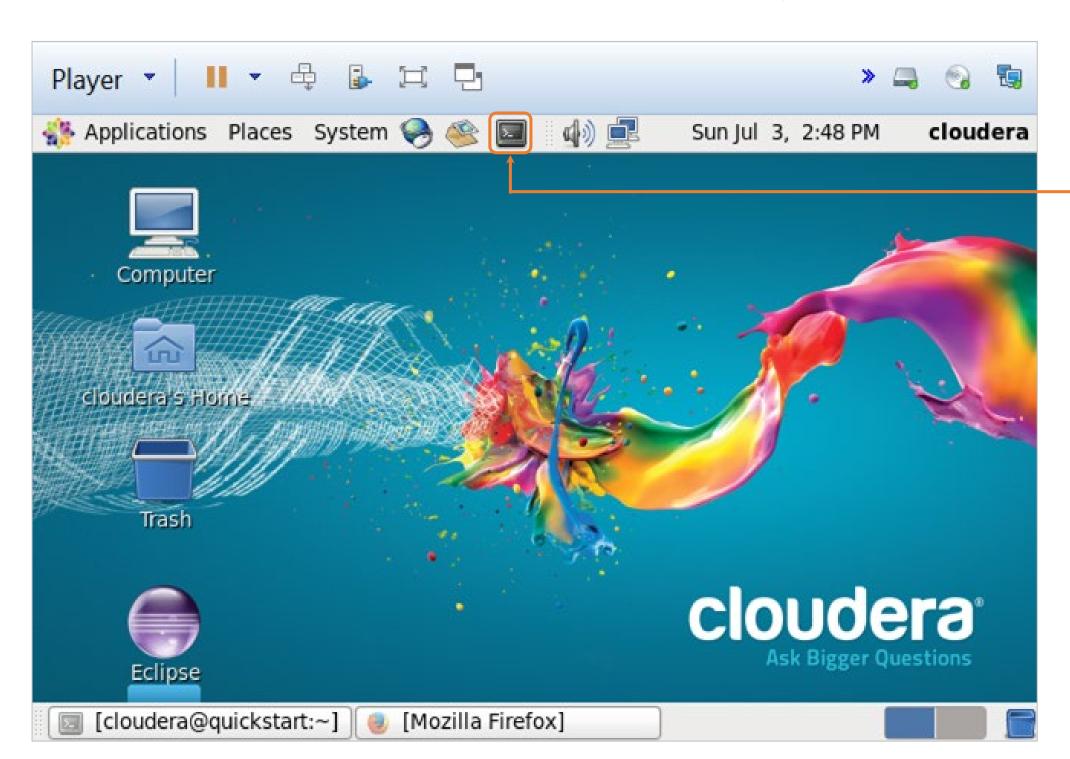
https://www.vmware.com/products/fusion/fusion-evaluation.html





Quick Start VM ware Image

Launch VMware player with Cloudera VM



Launch Terminal

Account:

username: cloudera

password: cloudera

QuickStart VM Terminal

Step 1



Step 2

```
cloudera@quickstart:~
                                                                                     _ D X
File Edit View Search Terminal Help
[cloudera@quickstart ~]$ pwd
/home/cloudera
[cloudera@quickstart ~]$ ls -lrt
total 1036
drwxrwsr-x 9 cloudera cloudera
                                4096 Feb 24 2013 eclipse
drwxrwxr-x 4 cloudera cloudera
                                4096 Apr 23 2015 workspace
drwxrwxr-x 2 cloudera cloudera
                                4096 Apr 23 2015 lib
drwxrwxr-x 4 cloudera cloudera
                                4096 Apr 23 2015 Documents
drwxrwxr-x 2 cloudera cloudera
                                4096 Apr 23 2015 Desktop
drwxrwxr-x 2 cloudera cloudera
                                4096 Apr 23 2015 datasets
-rw-rw-r-- 1 cloudera cloudera
                                1092 Apr 23 2015 cm api.sh
-rwxrwxr-x 1 cloudera cloudera
                                3978 Apr 23 2015 cloudera-manager
drwxr-xr-x 2 cloudera cloudera
                                4096 May 14 2015 Videos
                                4096 May 14 2015 Templates
drwxr-xr-x 2 cloudera cloudera
drwxr-xr-x 2 cloudera cloudera
                                4096 May 14 2015 Public
drwxr-xr-x 2 cloudera cloudera
                                4096 May 14 2015 Pictures
                                4096 May 14 2015 Music
drwxr-xr-x 2 cloudera cloudera
                                4096 May 14 2015 Downloads
drwxr-xr-x 2 cloudera cloudera
-rw-rw-r-- 1 cloudera cloudera 984565 Jun 30 12:00 test file
-rw-rw-r-- 1 cloudera cloudera
                                 187 Jun 30 12:04 mapper.py
                                  51 Jun 30 12:07 example test file
-rw-rw-r-- 1 cloudera cloudera
-rw-rw-r-- 1 cloudera cloudera
                                 868 Jun 30 12:16 reducer.py
-rw-rw-r-- 1 cloudera cloudera
                                  21 Jul 3 15:04 test 01
[cloudera@quickstart ~]$
```

Unix command:

- pwd to verify present working directory
- Is -lrt to list files and directories

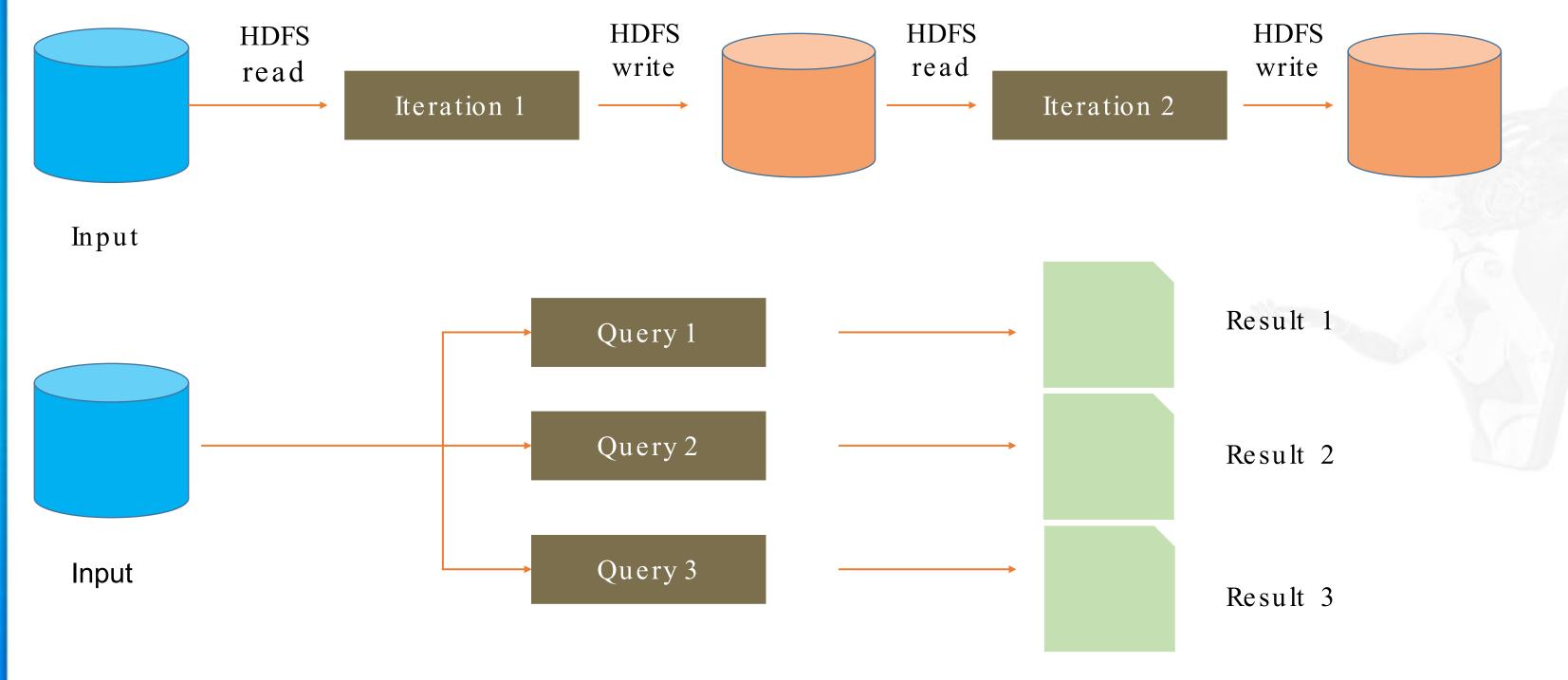


Using Hadoop Streaming for Calculating Word Count

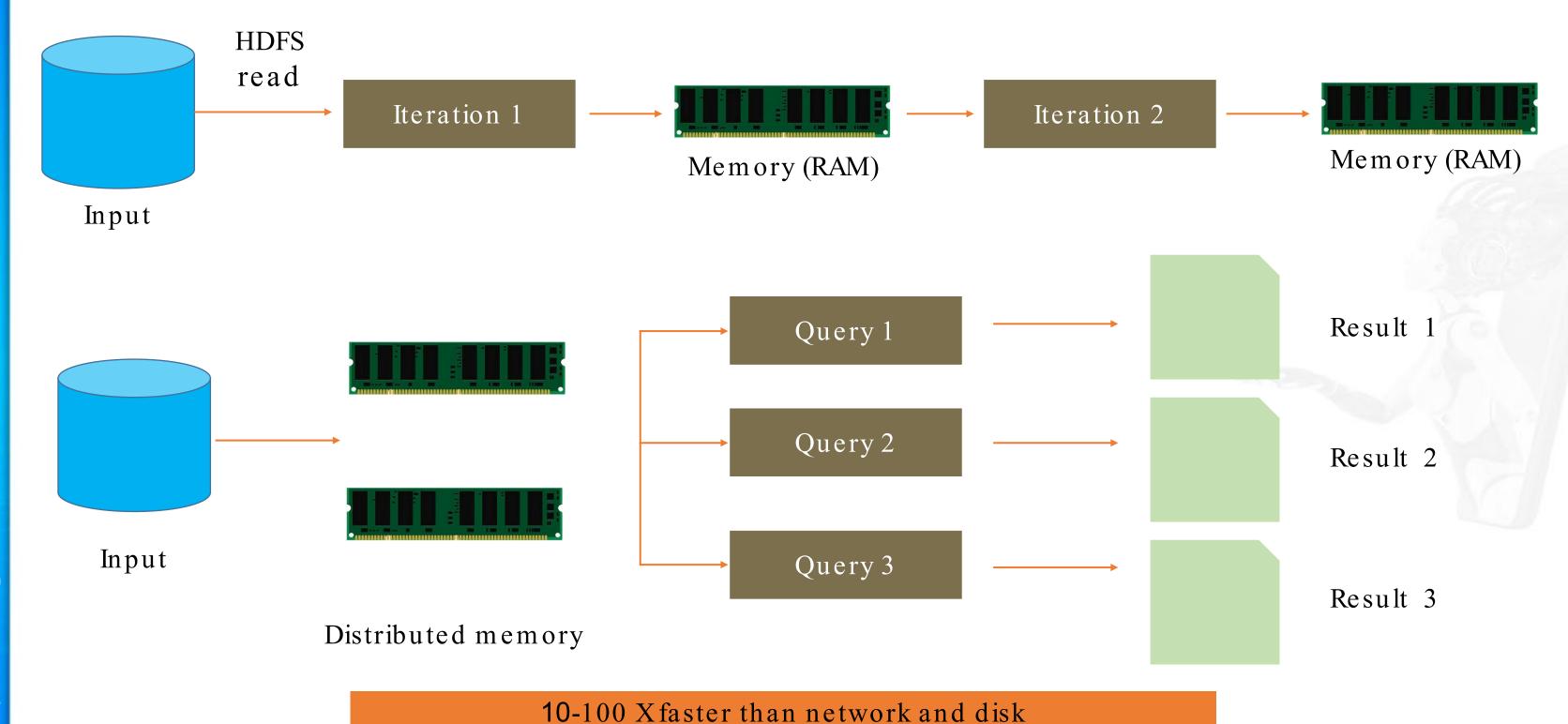


Demonstrate how to create a MapReduce program and use Hadoop Streaming to determine the word count of a document

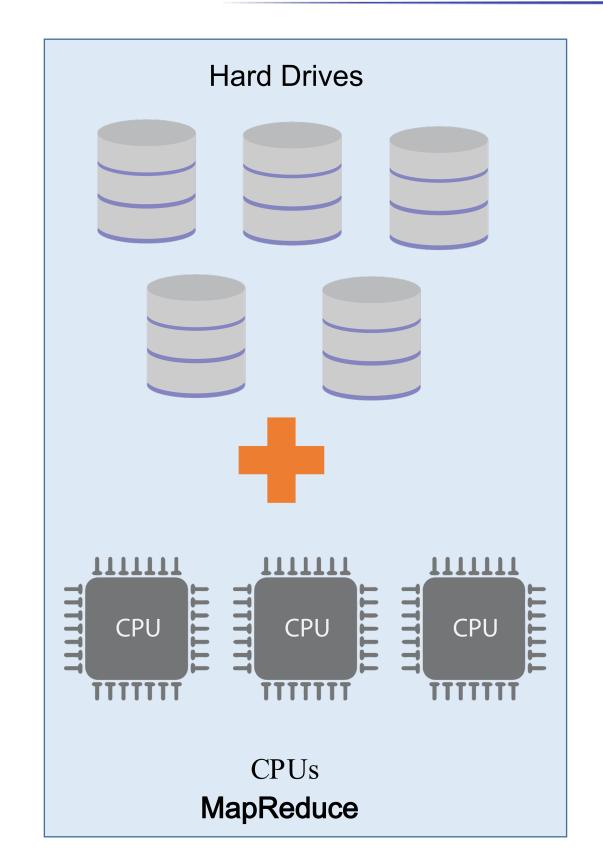
MapReduce Uses Disk I/O Operations

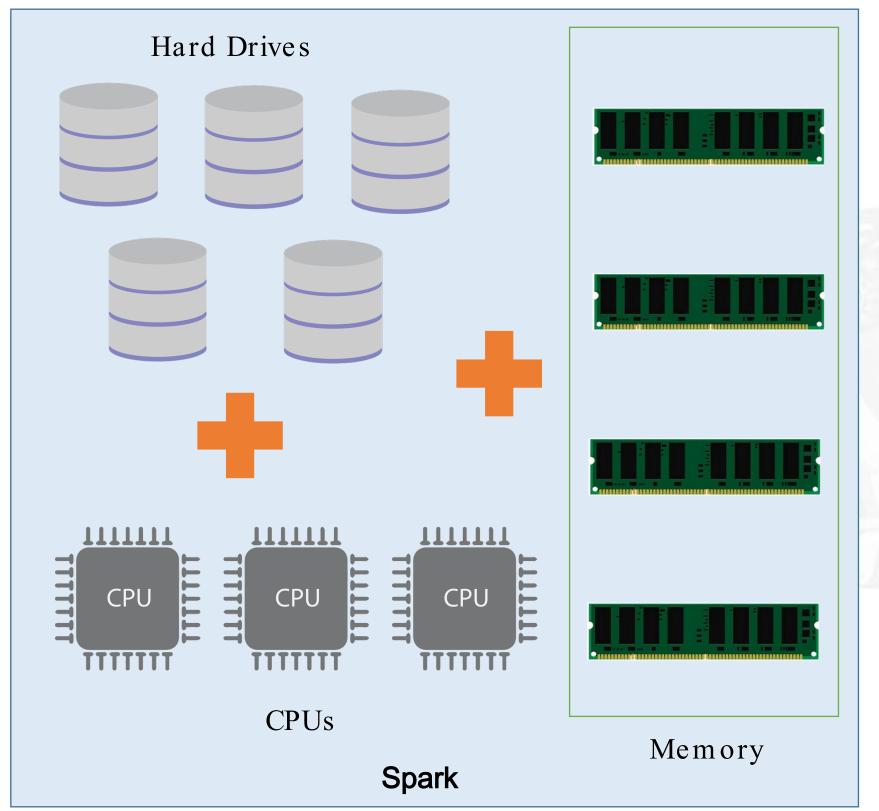


Apache Spark Uses In-Memory Instead of Disk I/O



Hardware Requirements for MapReduce and Spark





Apache Spark Resilient Distributed Systems (RDD)



Some basic concepts about Resilient Distributed Datasets (RDD) are:

- The main programming approach of Spark is RDD.
- They are fault-tolerant collections of objects spread across a cluster that you can operate on in parallel. They can automatically recover from machine failure.
- You can create an RDD either by copying the elements from an existing collection or by referencing a dataset stored externally.
- RDDs support two types of operations: transformations and actions.
 - Transformations use an existing dataset to create a new one.
 - Example: Map, filter, join
 - Actions compute on the dataset and return the value to the driver program.
 - Example: Reduce, count, collect, save



If the available memory is insufficient, then the data is written to disk.

Advantages of Spark



Faster: 10 to 100 times faster than Hadoop MapReduce

Simplified:

• Simple data processing framework

• Interactive APIs for Python for faster application development

Efficient:

Has multiple tools for complex analytics operations

Integrated:

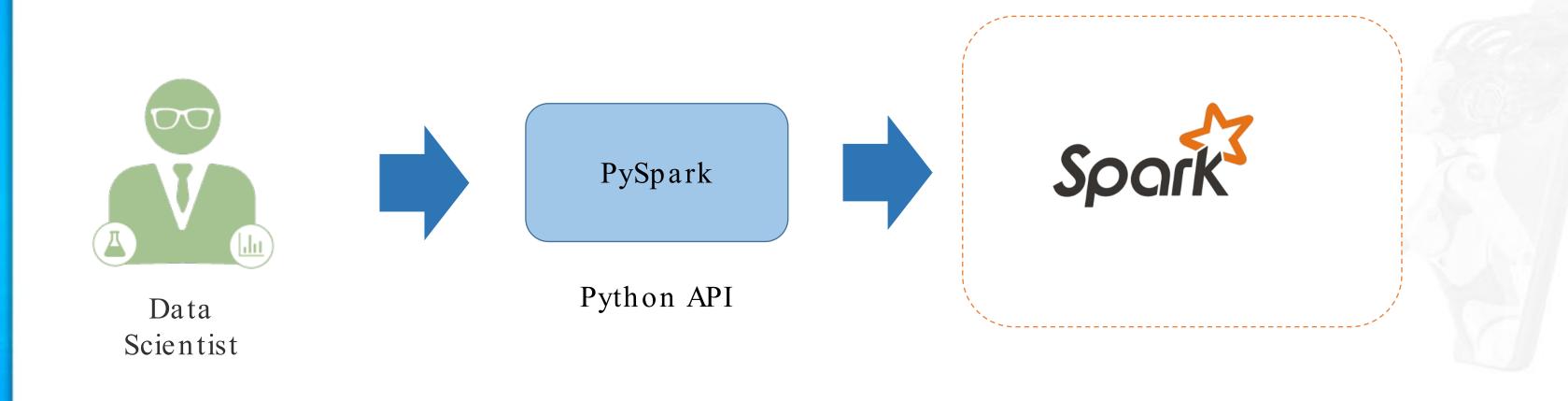
Can be easily integrated with existing Hadoop

infrastructure



PySpark: Python API for Spark

PySpark is the Spark Python API which enables data scientists to access Spark programming model.



PySpark: RDD Transformations and Actions

Transformation

Transformation	Description			
map()	Returns RDD, formed by passing data element of the source			
filter()	Returns RDD based on selection			
flatMap()	Maps items present in the dataset and returns sequence			
reduceByKey()	Returns key value pairs where values for which each key is aggregated by value			

Action

Action	Description		
collect()	Returns all elements of the dataset as an array		
count()	Returns the number of elements present in the dataset		
first()	Returns the first element in the dataset		
take(n)	Returns number of elements (n) as specified by the number in the parenthesis		

SparkContext or SC is the entry point to spark for the spark application

Spark Tools

Spark SQL

Spark Streaming MLlib (machine learning)

GraphX (graph)

Spark



Interactive Python APIs

Setting Up Apache Spark

To set up the Apache Spark environment, access the link:

http://spark.apache.org/downloads.html

Please use 7-Zip to extract these files.



Download	Libraries ▼	Documentation ▼	Examples	Community -	FAQ

Download Apache Spark™

Our latest stable version is Apache Spark 1.6.2, released on June 25, 2016 (release notes) (git tag)

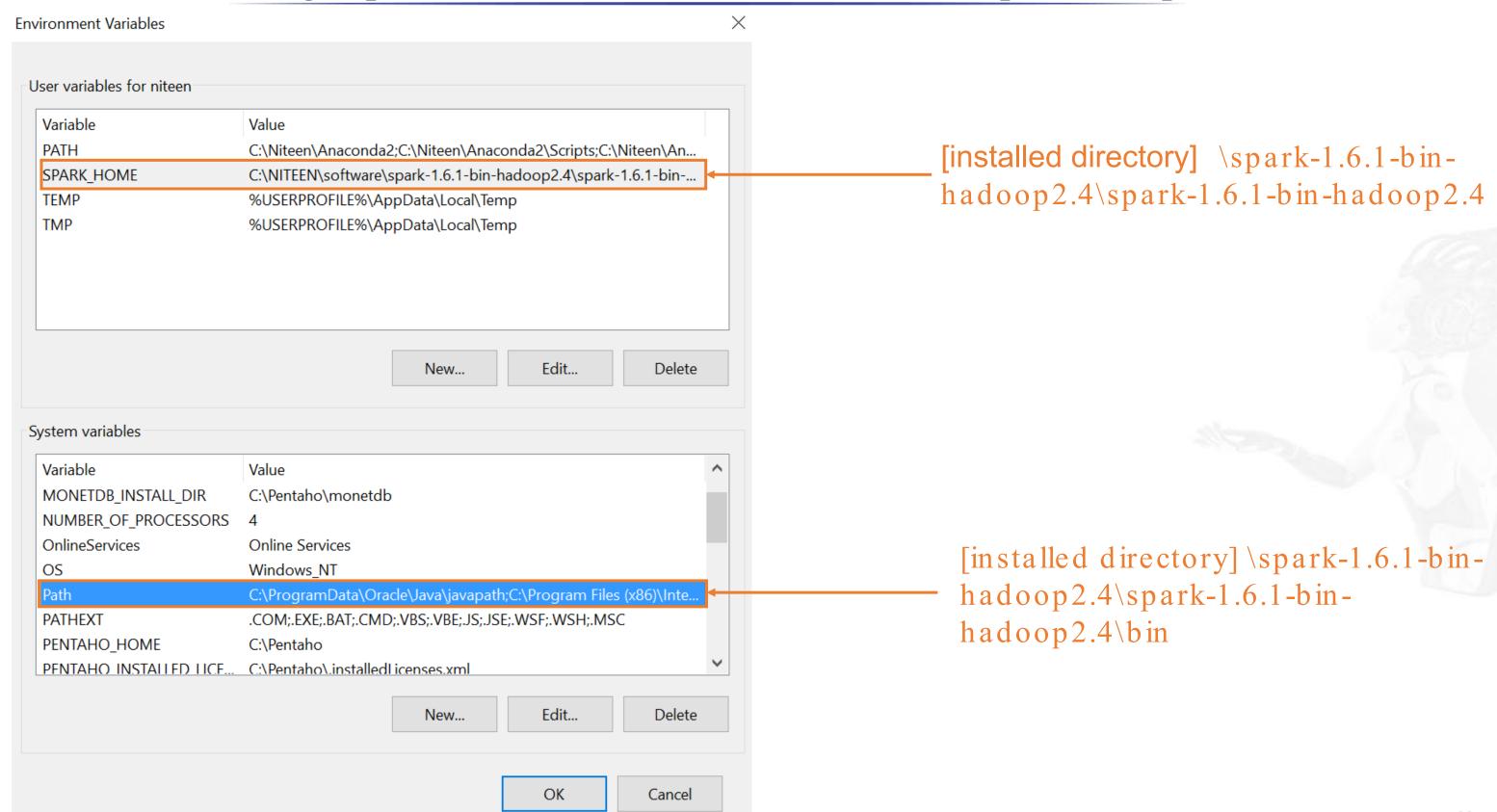
1. Choose a Spark release:	1.6.2 (Jun 25 2016)	•		
2. Choose a package type:	Pre-built for Hadoop 2.4			,
3. Choose a download type:	Direct Download ▼			
4. Download Spark: spark-1.	6.2-bin-hadoop2.4.tgz			

5. Verify this release using the 1.6.2 signatures and checksums.

Note: Scala 2.11 users should download the Spark source package and build with Scala 2.11 support.

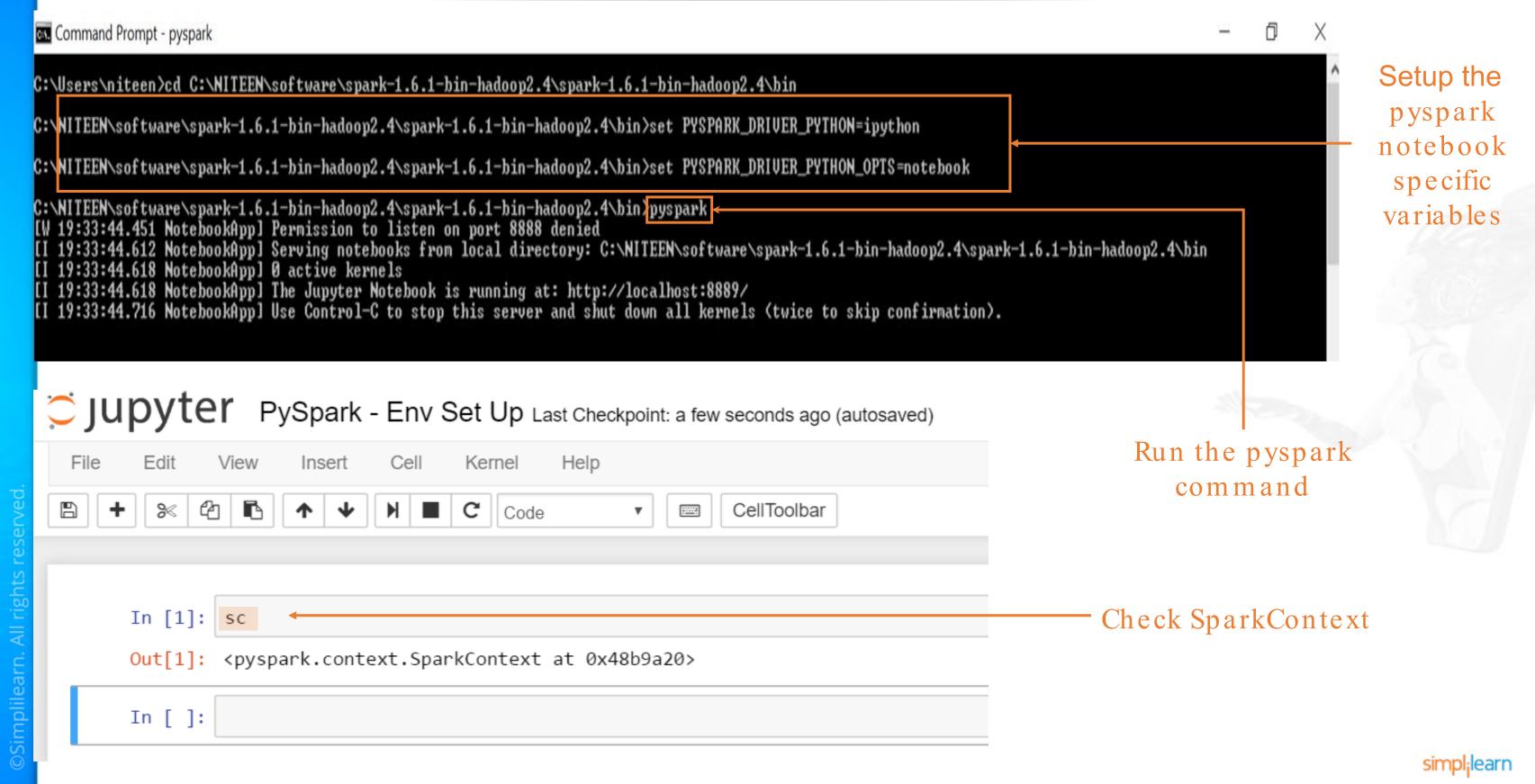


Setting Up Environmental Variable of Apache Spark



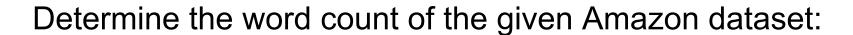


Integrating Jupyter Notebook with Apache Spark

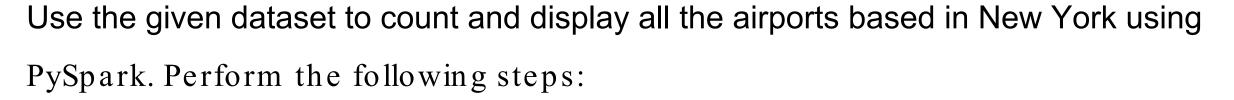


Using PySpark to Determine Word Count

Demonstrate how to use the Jupyter integrated PySpark API to determine the word count of a given dataset



- Create a MapReduce program to determine the word count of the Amazon dataset
- Submit the MapReduce task to HDFS and run it
- Verify the output



- View all the airports listed in the dataset
- View only the first 10 records
- Filter the data for all airports located in New York
- Clean up the dataset if required

DATA AND ARTIFICIAL INTELLIGENCE



Knowledge Check



What are the core components of Hadoop? Select all that apply.

- a. MapReduce
- b. HDFS
- c. Spark
- d. RDD





What are the core components of Hadoop? Select all that apply.

- a. MapReduce
- b. HDFS
- c. Spark
- d. RDD



The correct answer is **a and b**

MapReduce and HDFS are the core components of Hadoop.



2

MapReduce is a data processing framework which gets executed _____.

- a. at DataNode
- b. at NameNode
- c. on client side
- d. in memory





2

MapReduce is a data processing framework which gets executed _____.

- a. at DataNode
- b. at NameNode
- c. on client side
- d. in memory



The correct answer is a

The MapReduce program is executed at the data node and the output is written to the disk



3

Which of the following functions is responsible for consolidating the results produced by each of the Map() functions/tasks?

- a. Reducer
- b. Mapper
- c. Partitioner
- d. All of the above





3

Which of the following functions is responsible for consolidating the results produced by each of the Map() functions/tasks?

- a. Reducer
- b. Mapper
- c. Partitioner
- d. All of the above



The correct answer is a

Reducer combines or aggregates results produced by mappers.



4

What transforms input key -value pairs to a set of intermediate key -value pairs?

- a. Mapper
- b. Reducer
- c. Combiner
- d. Partitioner





4

What transforms input key -value pairs to a set of intermediate key -value pairs?

- a. Mapper
- b. Reducer
- c. Combiner
- d. Partitioner



The correct answer is a

Mapper processes input data to intermediate key -value pairs which are in turn processed by reducers.



Key Takeaways

You are now able to:

- Explain why Python should be integrated with Hadoop
- Outline the ecosystem and architecture of Hadoop
- Explain the functioning of MapReduce
- Discuss Apache Spark functions and their benefits
- Write Python programs for Hadoop operations



Stock Market Data Analysis



Import the financial data using Yahoo data reader for the following companies:

- Yahoo
- Apple
- Amazon
- Microsoft
- Google

Perform fundamental data analysis

- Fetch the previous year's data
- View the values of Apple's stock
- Display the plot of closing price
- Display the stock trade by volume
- Plot all companies' data together for closing prices

Stock Market Data Analysis



Perform Daily Return Analysis and show the relationship between different stocks

- Plot the percentage change plot for Apple's stock
- Show a joint plot for Apple and Google
- Use PairPlot to show the correlation between all the stocks

Perform risk analysis



On April 15, 1912, the Titanic sank after colliding with an iceberg, killing 1502 out of 2224 passengers and crew. This tragedy shocked the world and led to better safety regulations for ships. Here, we ask you to perform the analysis through the exploratory data analysis technique. In particular, we want you to apply the tools of machine learning to predict the survived passengers.

Titanic Data Set Analysis



The details of these projects and their scope are below:

Data acquisition of the Titanic dataset

- Train dataset
- Test dataset

Perform the Exploratory Data Analysis (EDA) for train dataset

- Passengers age distribution
- Passengers survival by age
- Passengers survival breakdown
- Passengers class distribution
- Passengers embarkation by locations



Perform machine learning to train the model and

- Create user defined function to load train data set
- Create user defined function to load test data set
- Create machine model
- Train the machine
- Predict whether a passenger survived the tragedy or not
- Persist the mode for future re -use



Thank You

