

# Distributed Processing: MapReduce Framework ©Simplilearn. All rights reserved. simpl<sub>i</sub>learn

# Introduction to MapReduce



- MapReduce is a programming model that simultaneously processes and analyzes huge data sets logically into separate clusters.
- Map sorts the data, whereas Reduce segregates it into logical clusters, thus removing the bad data and retaining the necessary information.

# **Learning Objectives**

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

- Perform distributed processing in MapReduce
- Illustrate how MapReduce deals with distributed processing issues
- Control the flow of mappers and reducers
- Optimize MapReduce jobs





Distributed Processing in MapReduce



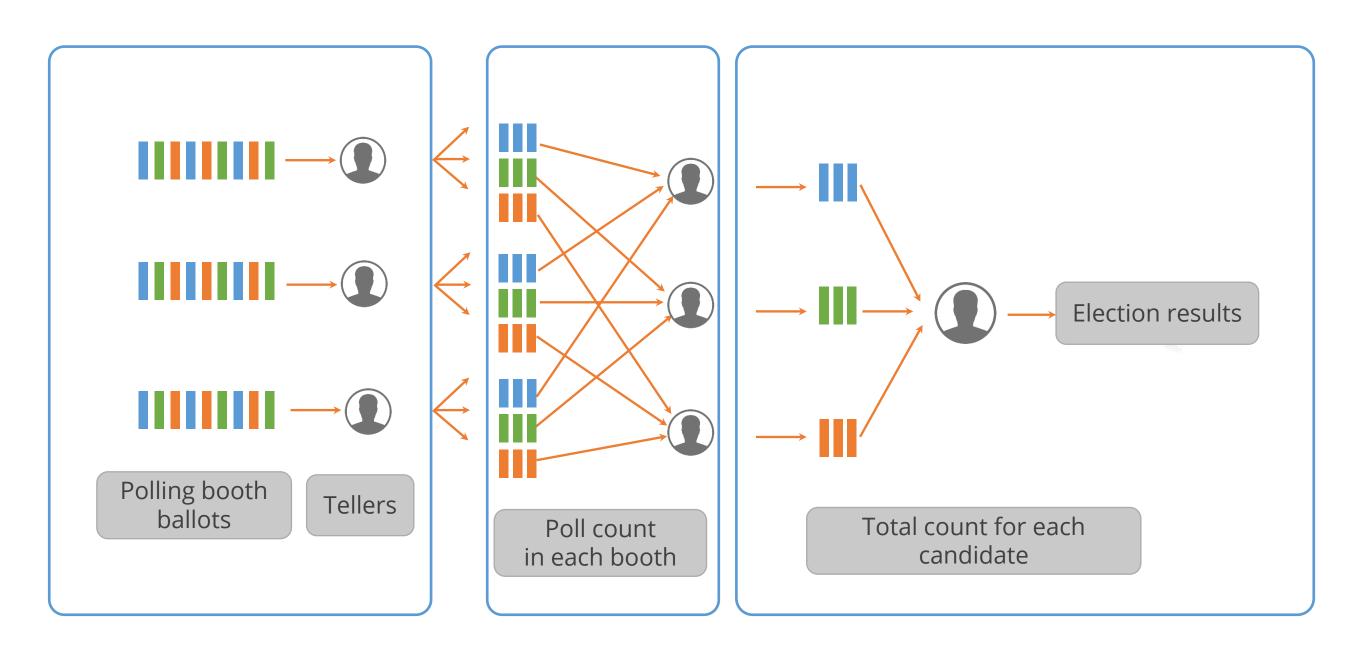
# Why MapReduce?



Huge amounts of data were stored on single servers before 2004.

# **MapReduce: Analogy**

Here is a diagram for an analogy of how MapReduce works.

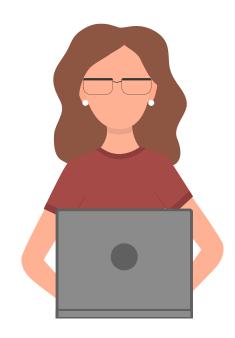


# **MapReduce: Analogy**

Here is an example of election results to understand how MapReduce works.



One month to receive the election results



Individual work



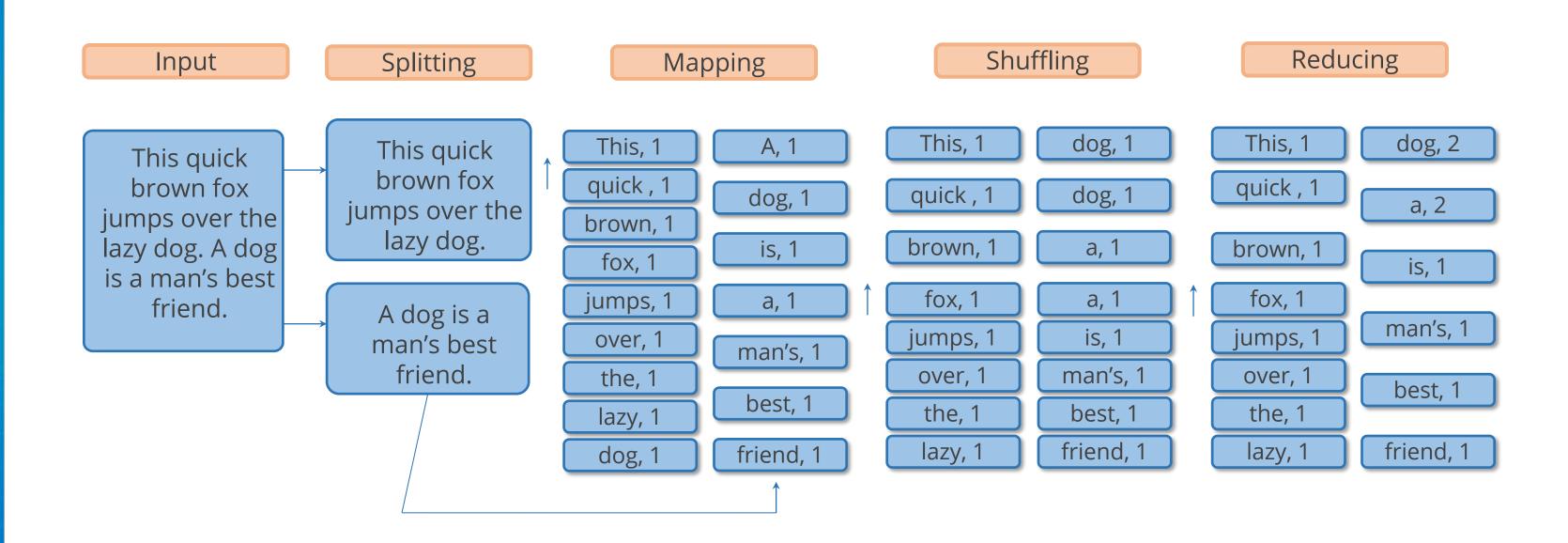
Results obtained in one or two days



Parallel work

# **MapReduce: Word Count Example**

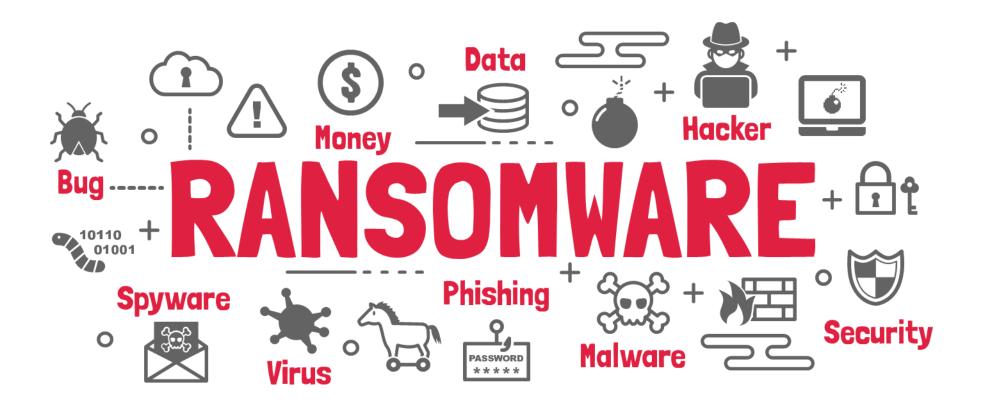
The overall MapReduce word count process is shown below:



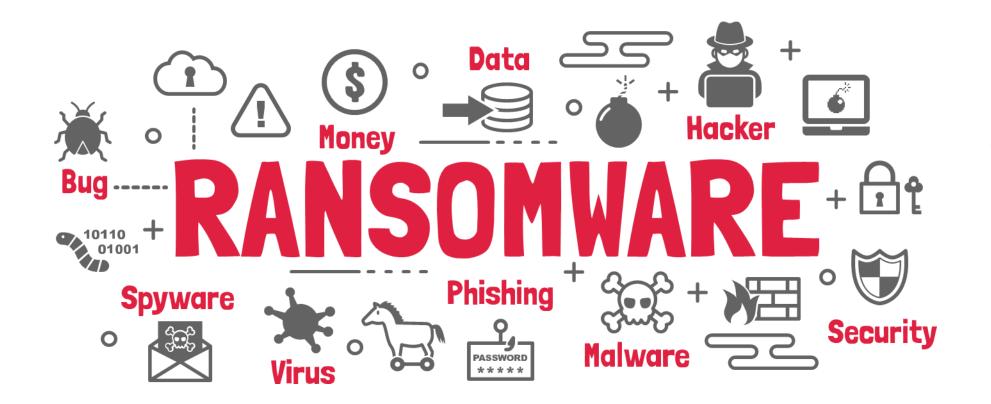




Ransomware caused a stir by infiltrating Windows systems and encrypting data. To recover the data, the owner or organization was forced to pay a ransom in Bitcoin.



The WannaCry ransomware was rapidly spreading across systems in 2018, particularly on Windowsbased OS. Flipkart (backed by Walmart) ran 90 percent of its servers on Windows, increasing the probability of an attack.





### **Challenges**

- Flipkart managed to acquire a security patch from Microsoft to fix the vulnerabilities. However, the main issue was how and when to apply the patch to the systems.
- There were almost 10,000 servers that needed to be patched. The biggest question was at what time the patching should happen.



#### **Solution**

- Server logs were analyzed to find the precise time when traffic on the servers was low and the patch could be applied.
- MapReduce, a Hadoop component, was selected to analyze and process server logs.

Using MapReduce on 200 terabytes of data, Flipkart found that the request count was lowest at 4 a.m. every day and decided to shut down the servers to apply the Microsoft patch.









#### Keys

- It is a recording entity received at the mapper given by RecordReader.
- For example, in (country,1), country is a key. Keys are driven by use case and are selected by users.

#### Value

- It is a part of a recording entity and holds data for computing.
- For example, in (country,1), 1 is a value.

#### InputSplit

- It's a logical representation of data that InputFormat generates and maps to a single task that a mapper will execute.
- It describes a unit of work that contains a single map task in MapReduce.



#### RecordReader

• It is an internal framework that converts the byte-oriented view of data from InputSplit to key-value pairs and passes them on to the Map.

#### Mapper

- It is a self-sustained function that takes input from RecordReader and processes it.
- The input for Mapper is a Map (K, V), and the output is a Map (K1, V1). It is stored on HDFS in the temp directory.

#### Reducer

- It is a function that takes input from the Mapper output.
- Then, it performs computations, such as sum, filter, and aggregation, on them and provides a Map (K2, V2) as an output.



#### **Partitioner**

- It decides the partitioning of keys for the intermediate map outputs.
- It also controls the middle map output segment.

#### **Counter**

- It is a function to gather statistics about the MapReduce job.
- These are inbuilt but can be user-defined as well.

#### **Combiner**

- It is also known as a mini-reducer.
- It works at the mapper level and optimizes data transfer to reduce congestion before passing it to the reducer.

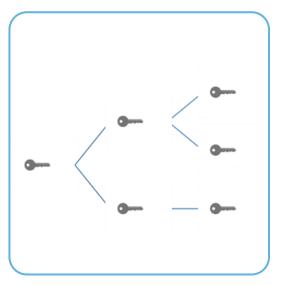


# **MapReduce Essentials**

The job input is specified in key-value pairs:



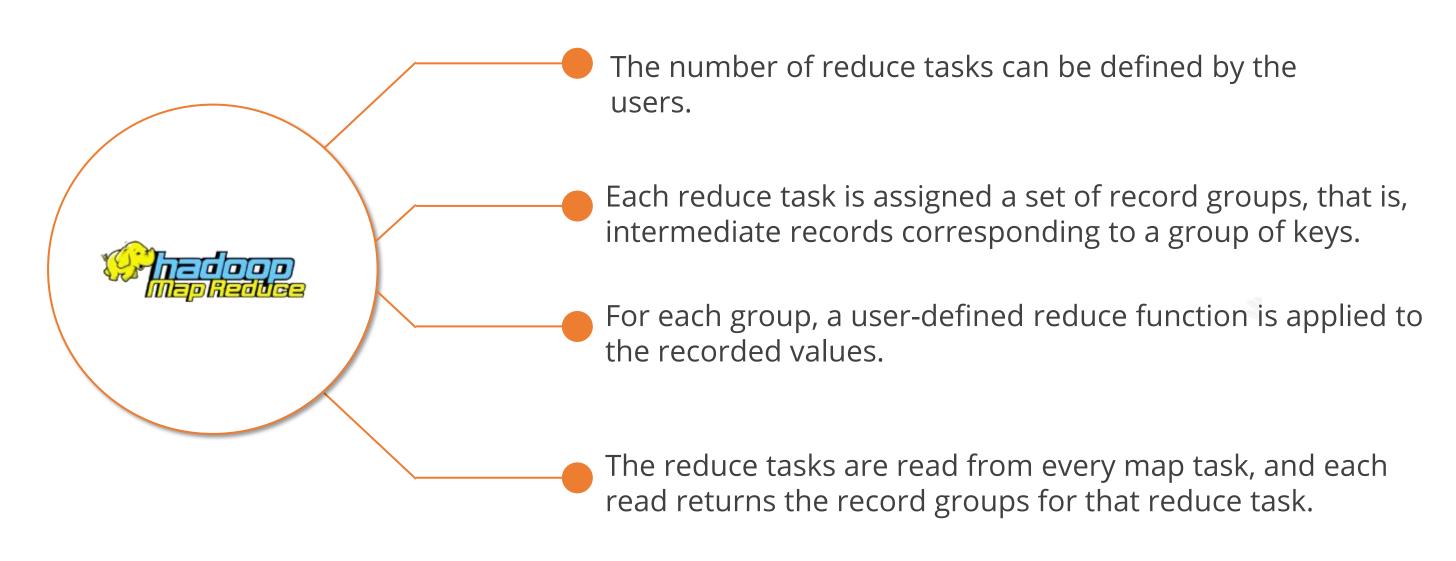
A user-defined map function is applied to each input record to produce a list of intermediate keyvalue pairs.



A user-defined reduce function is called once for each distinct key in the map output.

## **MapReduce Essentials**

The essential steps of each MapReduce phase include:



Reduce phase cannot start until all mappers have finished processing.





**Map Execution Phases** 



# **Map Execution Phases**



#### Map phase

- Reads assigned input split from **HDFS**
- Parses input into records as keyvalue pairs
- Applies map function to each record
- Informs master node of its completion



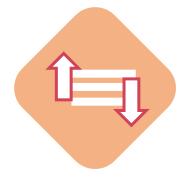
#### **Partition phase**

- Each mapper must determine which reducer will receive each of the outputs
- For any key, the destination partition is the same
- Number of partitions is equal to the number of reducers



**Shuffle phase** 

 Fetches input data from all map tasks for the portion corresponding to the reduce tasks' bucket



**Sort phase** 

 Merge sorts all map outputs into a single run



#### **Reduce phase**

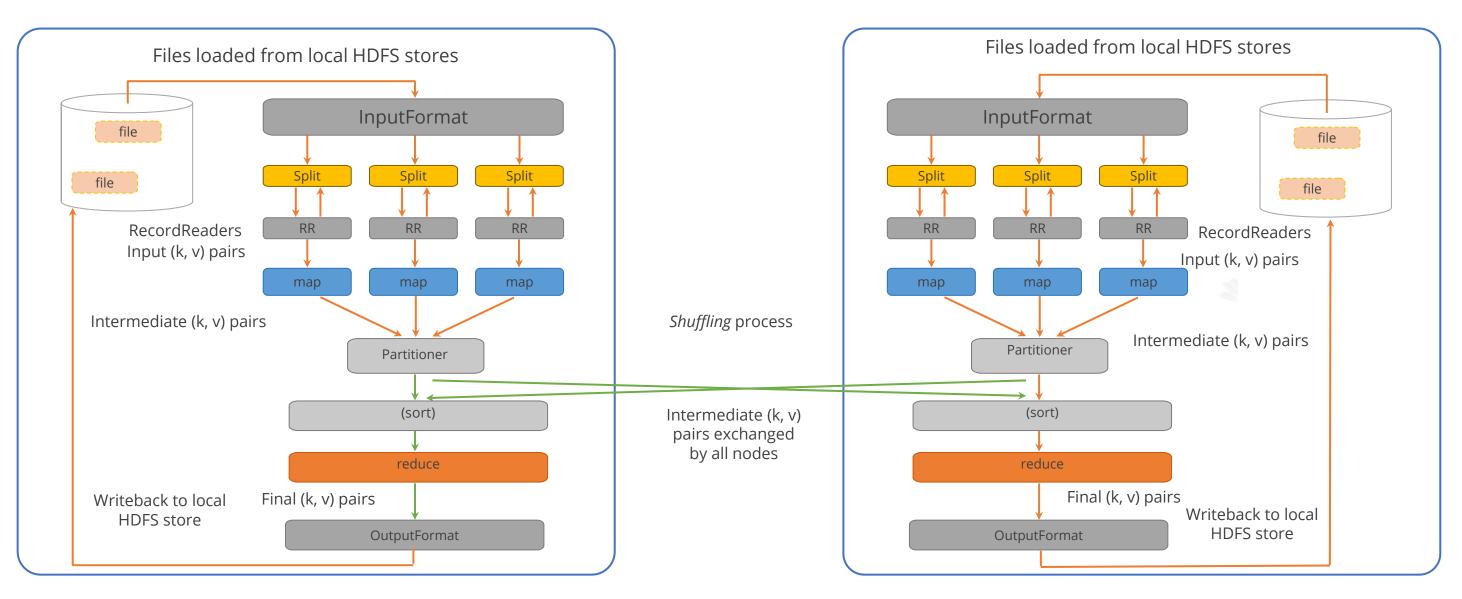
 Applies userdefined reduce function to the merged run



## **Map Execution: Distributed Two-Node Environment**

Here is a diagram representing a two-node distributed environment.

Node 1 Node 2







MapReduce Jobs



## **MapReduce Jobs**

A job is a MapReduce program that causes multiple map and reduce functions to run parallelly over the life of the program.

ApplicationMaster and NodeManager functions:

#### ApplicationMaster

- Is responsible for the execution of a single application or MapReduce job
- Divides job requests into tasks and assigns them to NodeManagers running on the slave node



#### NodeManager

- Has many dynamic resource containers
- Executes each active map or reduce task regularly with the ApplicationMaster

## **MapReduce and Associated Tasks**

#### Map process

Is an initial ingestion and transformation step where initial input records are processed parallely



ApplicationMaster
Keeps track of a MapReduce job

Is an aggregation

Is an aggregation or summarization step in which all associated records are processed together

**NodeManager** 

Keeps track of individual map tasks and can run in parallel to other NodeManager



## **Hadoop MapReduce Job Work Interaction**

Hadoop MapReduce

> Job submission

Hadoop MapReduce job is submitted by a client in the form of an input file or a number of input split of files containing data.

Input split distribution

MapReduce ApplicationMaster distributes the input split to separate NodeManagers.

Success or failure status

ResourceManager gathers the final output and informs the client of the success or failure status.

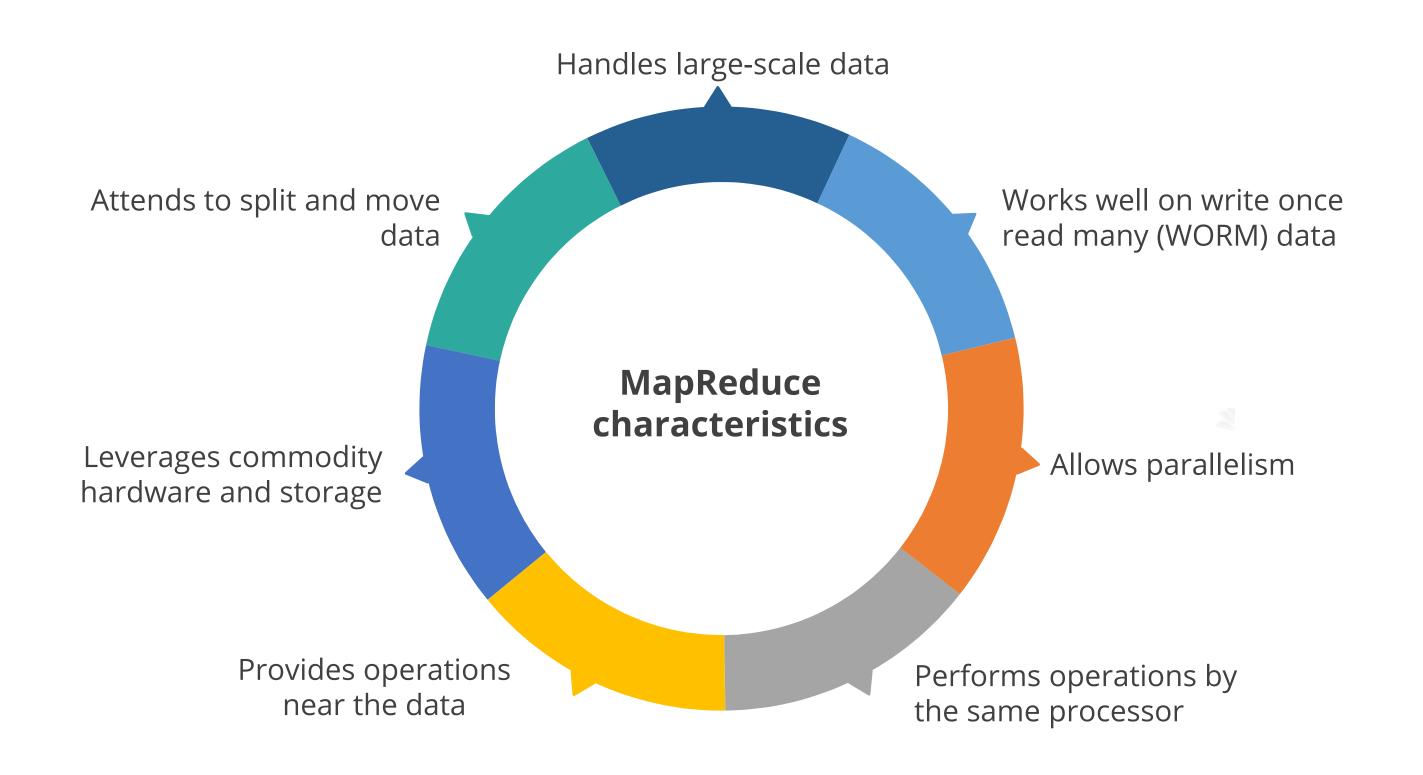
Resubmission of tasks

MapReduce ApplicationMaster resubmits the tasks to an alternate NodeManager if the DataNode fails.

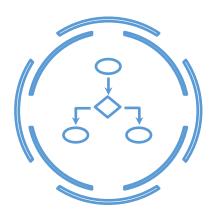
Coordination with NodeManager

MapReduce ApplicationMaster coordinates with the NodeManagers.

# **Characteristics of MapReduce**



# **Real-Time Uses of MapReduce**









Algorithms

Sorting

Data mining

Search engine operations







Gaussian analysis



Semantic web 3.0

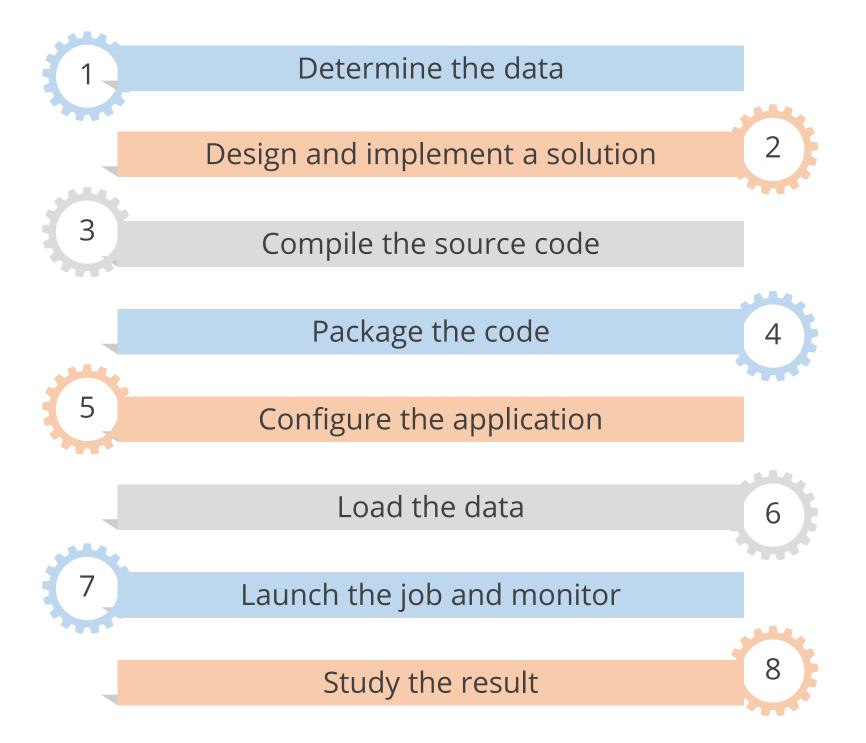


**Building a MapReduce Program** 



# **Building a MapReduce Program**

The steps to build a MapReduce program are:



## **Hadoop MapReduce Requirements**

The user or developer is required to set the framework with the following parameters:

Locations of the job input

Locations of the job output

Input format

Output format

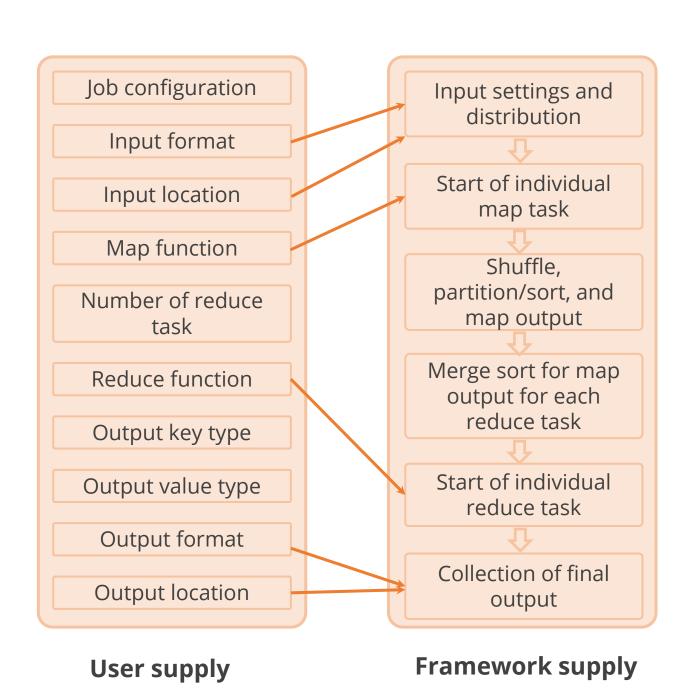
Class containing the map function

Class containing the reduce function



#### **Set of Classes**

The image below shows the set of classes under user supply and the framework supply.



- ResourceManager accepts the input and hands over the job to ApplicationMaster, which divides the job into tasks.
- NodeManager completes the assignment by executing the map task as part of container execution.
- The reducer starts the merging process.
- The output is collected.

# **MapReduce - Responsibilities**

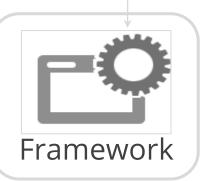
#### Responsibilities



Setting up the job

Specifying the input location

Ensuring correct format and location of input



Distributing the job

Running the map operation

Performing shuffling and sorting

Reducing phases

Placing the output

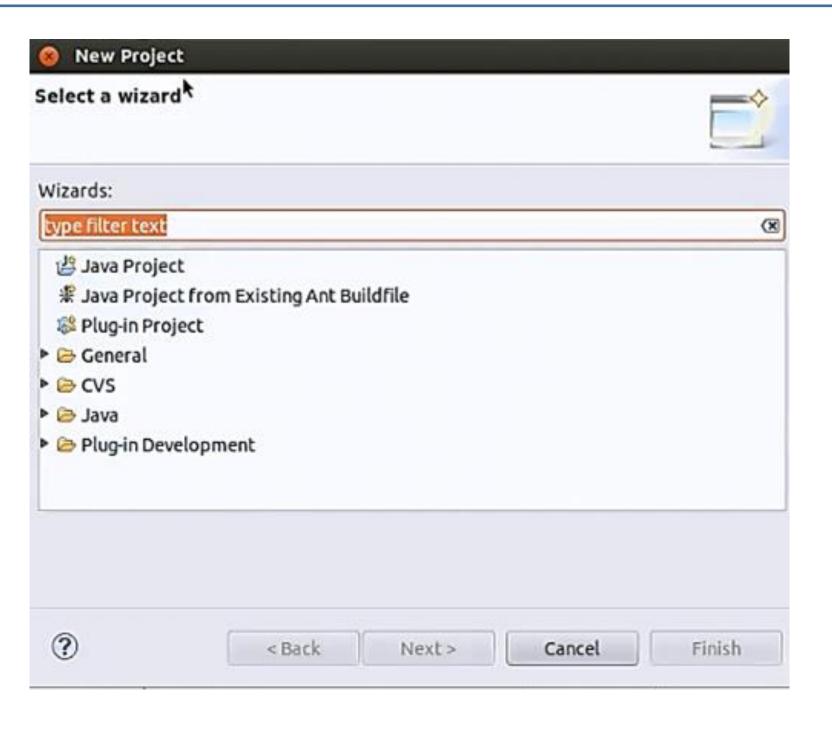
Informing the user of job completion status



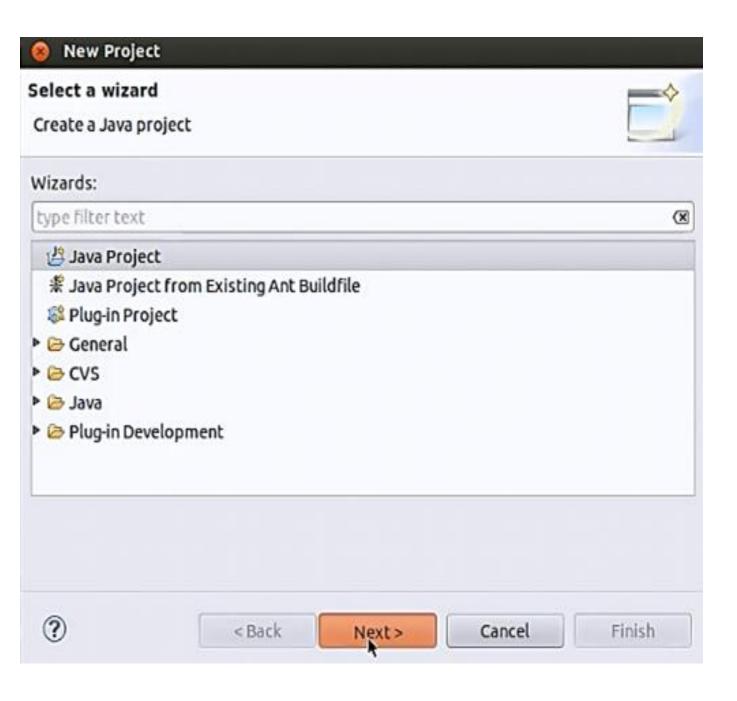
Creating a New Project on MapReduce



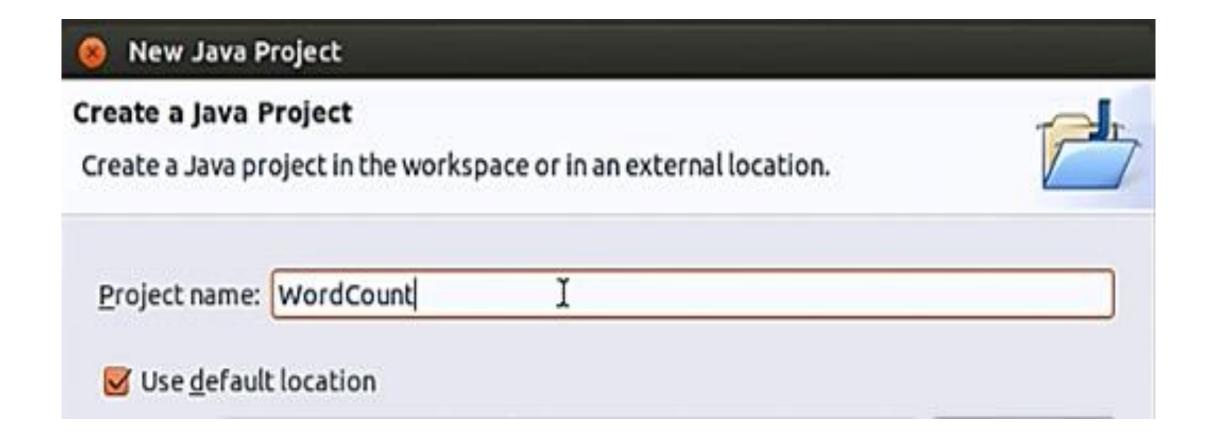
Create a new project and add essential JAR files to run MapReduce programs



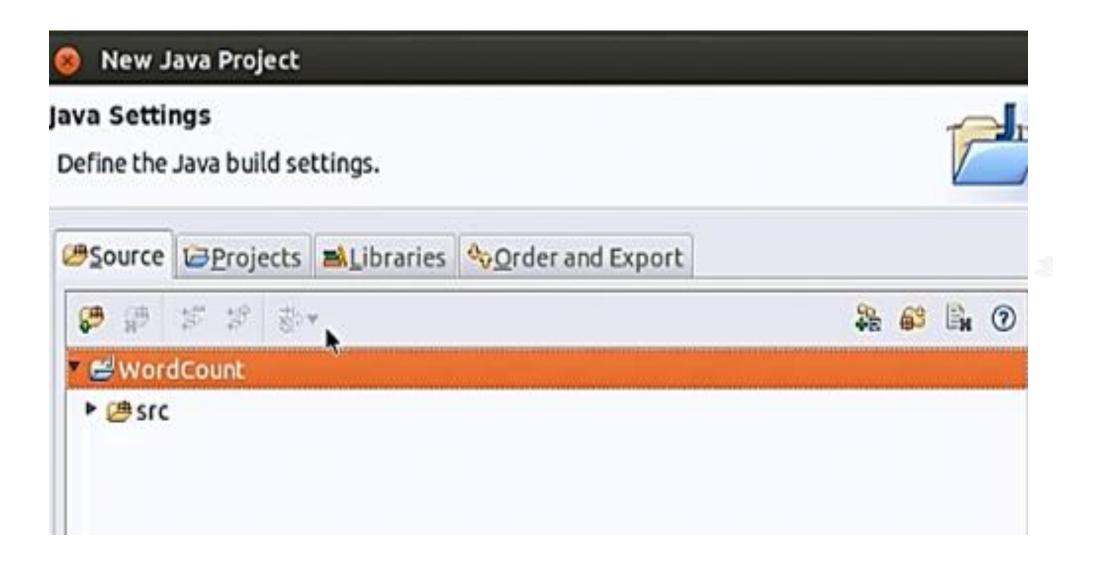
Select Java Project, and then click the Next



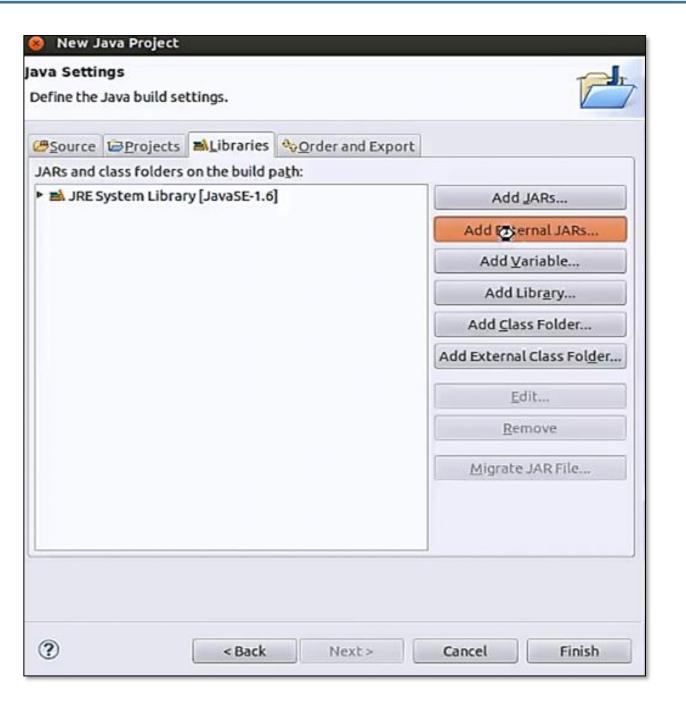
Enter the project name



Include JAR files from the Hadoop framework to ensure that the programs locate the dependencies in one location

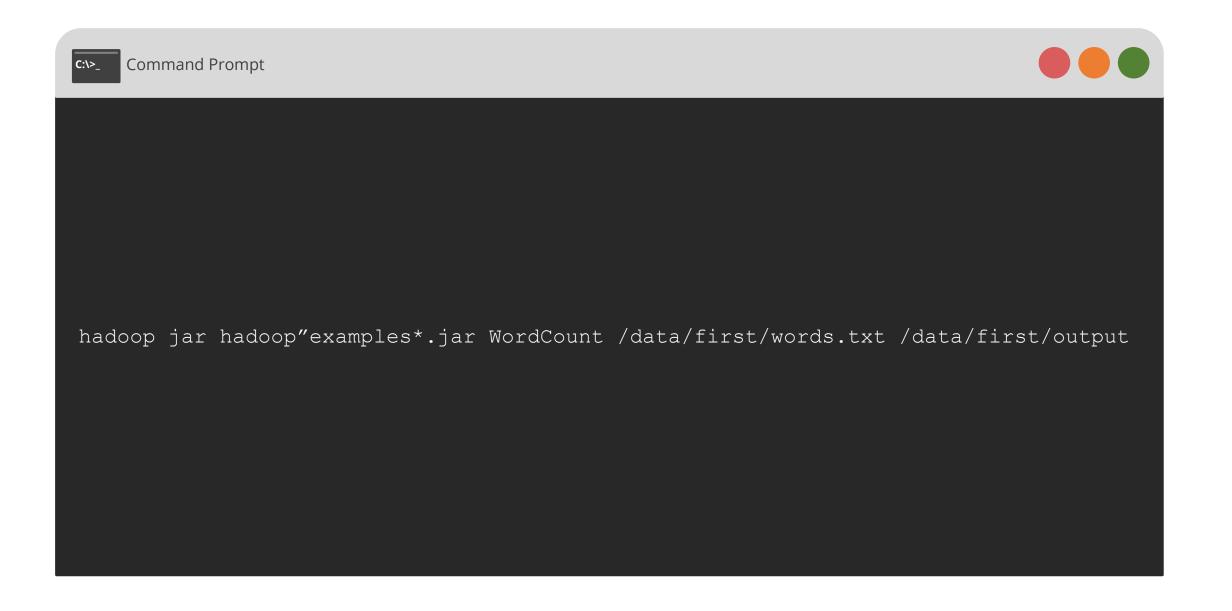


#### Add the essential JAR files



# **Checking Hadoop Environment for MapReduce**

Ensure that the machine setup can perform MapReduce operations



# **Controlling the Flow of Mapper and Reducer**

#### **Number of Maps**

- The number of maps is dependent on the total number of blocks in the input dataset.
- Having 10 to 100 maps per node is a common rule.

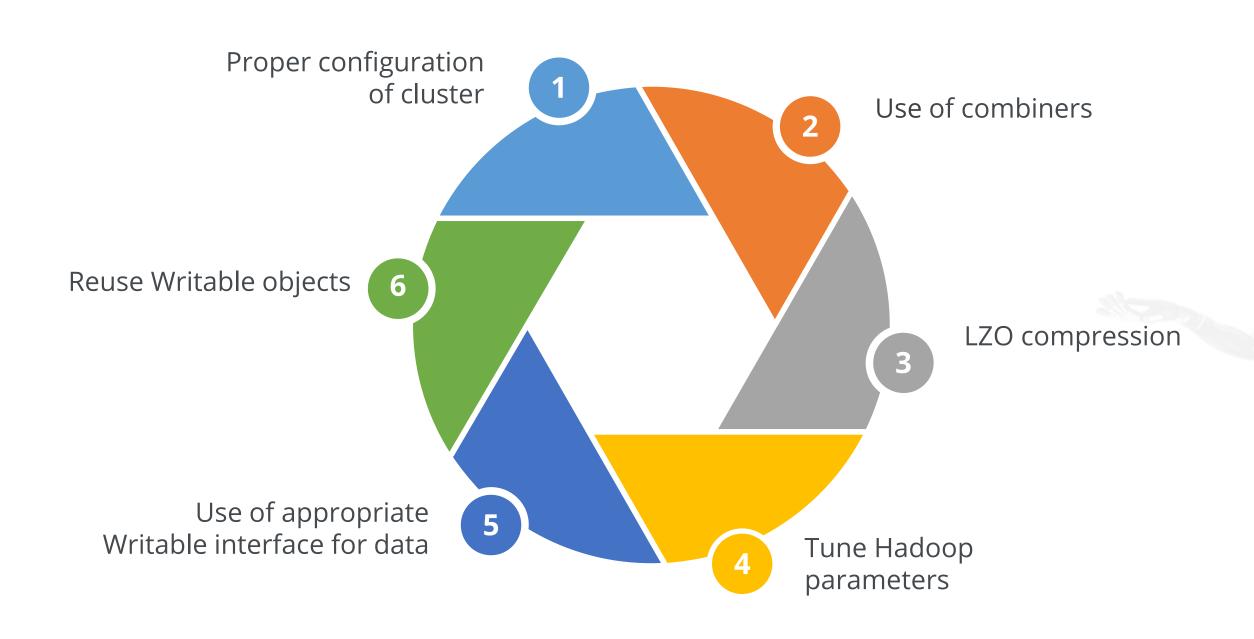
#### **Number of Reducers**

- The reducer uses a formula that multiplies 0.95 or 1.75 by the product of the number of nodes and the maximum number of containers per node.
- 0.95\* (number of nodes multiplied by the maximum number of containers per node)
- The reducer can also be set to none.

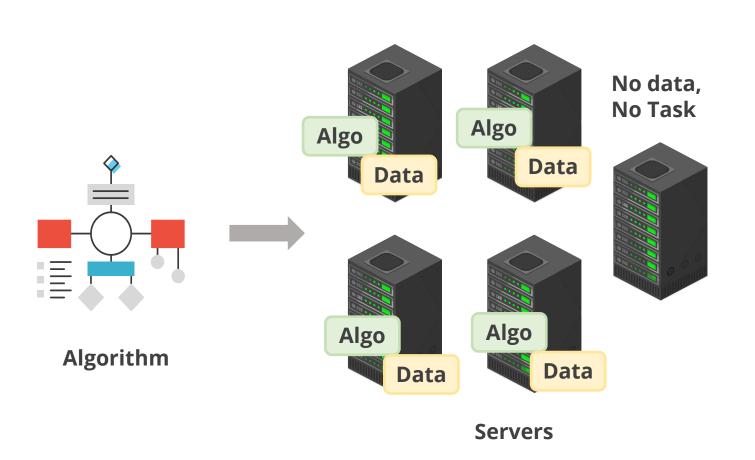


# **Optimization of MapReduce Jobs**

Follow the techniques to achieve maximum performance from the MR cluster:



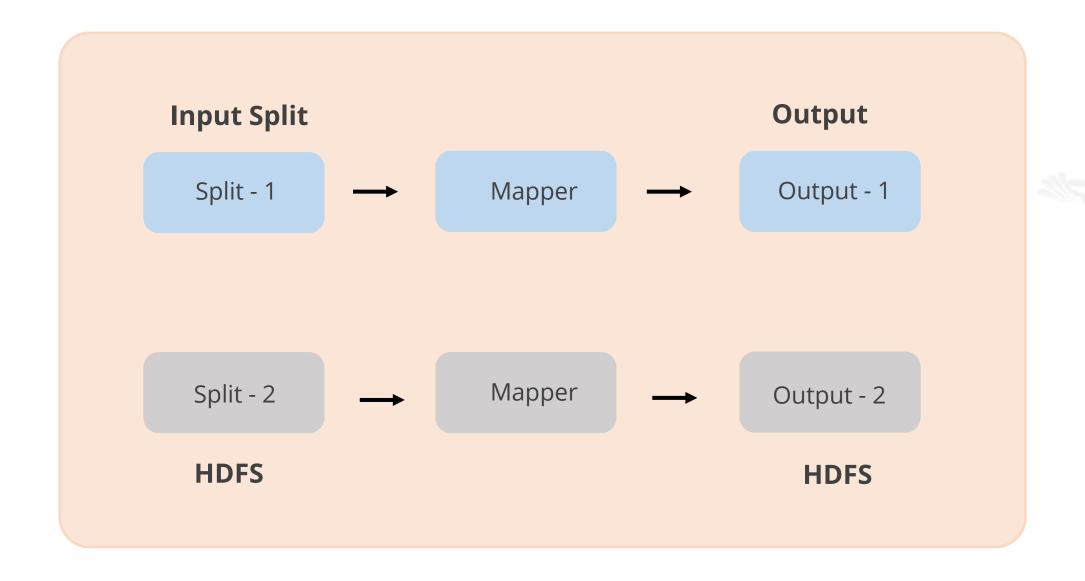
# **Fault-Tolerance and Data Locality**



- Fault tolerance refers to the system's capacity to continue running without interruption when one or more of a system's components fail.
- Data locality is a way to move job computation close to where the actual data resides instead of moving a large amount data to computation.
- It helps to minimize the overall network congestion and increases the overall throughput of the system.
- As we can see in the graphic, the data has not been distributed in the last node. When MR runs, it ignores the last node and no task is sent. Hence, there is no output from that node.

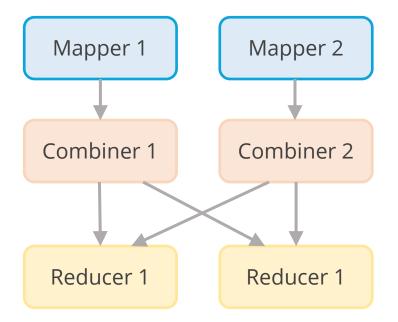
# **Working with Map-Only Jobs**

It is a process in which the mapper executes the tasks without a reducer. This performs well when one transformation is required without shuffling. Hence, the performance improves.



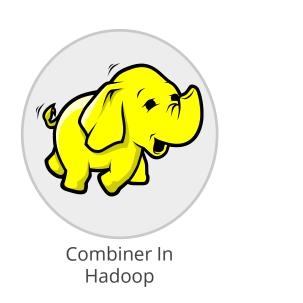
## **Combiners in MapReduce**

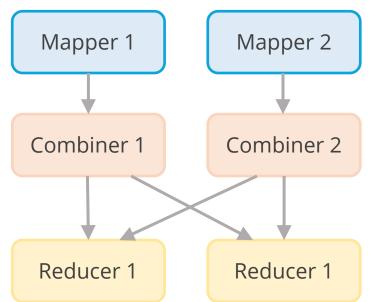




- Combiners are local Reducer.
- It reduces network congestion by acting as a bridge between the Mapper and Reducer.
- The output produced by the Mapper is the intermediate output in terms of key-value pairs.
- If the output is large, it cannot be directly fed to the Reducer as it would cause Network Congestion.

# **Combiners in MapReduce**

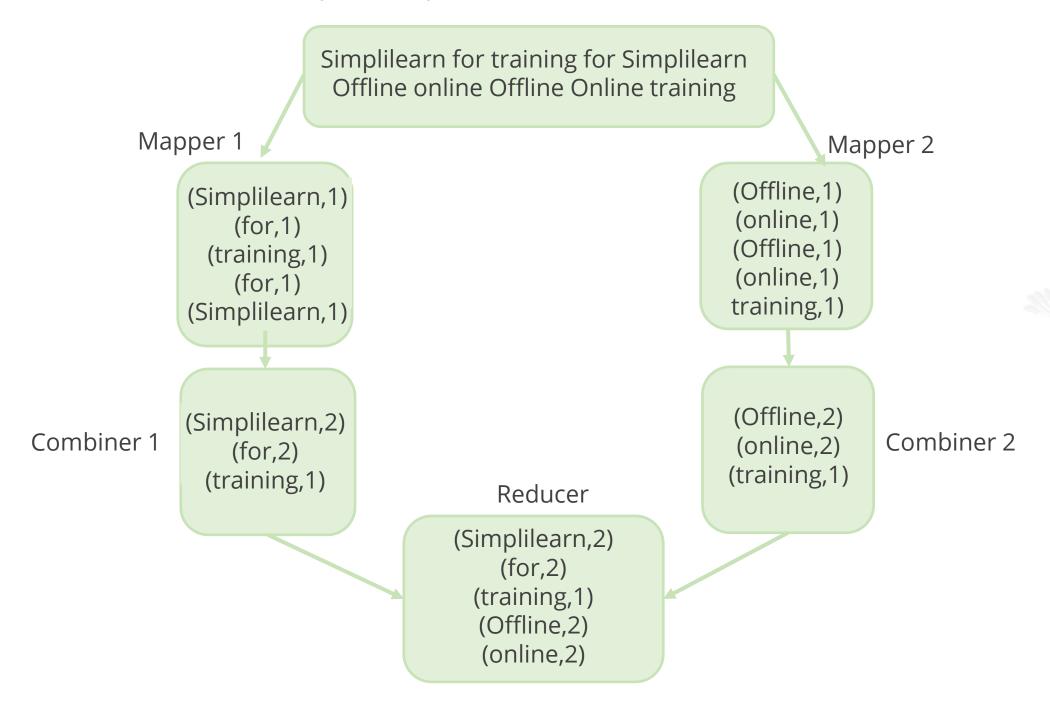




- A Combiner is used to minimize congestion.
- It is also known as a semi-reducer.
- It is optional in the MapReduce program.
- It provides a summary of large datasets.
- It improves the overall performance while dealing with large datasets.

# **MapReduce Program with Combiner**

In a program, the combiner accepts inputs from the Map class and then passes the output key-value pairs to the Reducer class.



# **Assisted Practice: Execution of MapReduce Job**



**Duration:** 10 mins

**Problem Scenario:** Write the commands to perform partitions using the JAR files

**Objective:** In this demonstration, you will use JAR files and the wordcount.txt file to perform the partitions using the MapReduce job.

**Dataset Name: "wordcount.txt"** 

# **Assisted Practice: Execution of MapReduce Job**



#### Tasks to Perform:

Step 1: Download the Hadoop-mapreduce-example.jar file and wordcount.txt file

Step 2: Log in to the FTP using the username and password from the lab and upload the file

Step 3: Log in to the Web Console using the username and password from the lab and create a new directory **demo** in HDFS using the **mkdir** command

Step 4: Push the wordcount.txt file into the directory using the put command

Step 5: Execute the command to move the **Hadoop-mapreduce-example.jar** file to the HDFS directory

Step 6: View the files in the **Output** folder with the part files

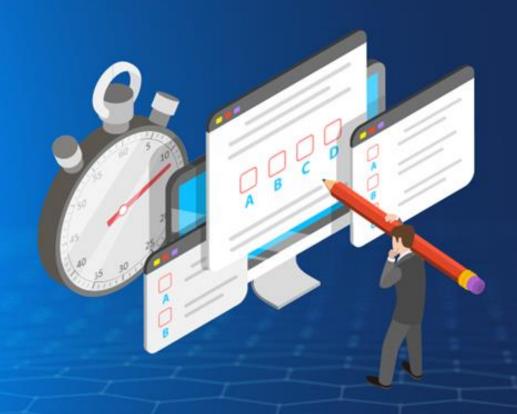
Note: The solution to this assisted practice is provided under the course resources section.

# **Key Takeaways**

- MapReduce divides and analyzes large data sets into logical clusters.
- A job is a MapReduce program that generates multiple maps and reduces functions throughout the program's lifetime.
- Tested techniques must be used to achieve maximum performance from the MR cluster.
- The MapReduce job divides the input dataset into separate chunks, which are then processed in parallel by the map jobs.



# DATA AND ARTIFICIAL INTELLIGENCE



**Knowledge Check** 



#### Which of the following defines the data locality feature of MapReduce?

- A. Reduce task in MapReduce is performed using the map() function
- B. Reduce task in MapReduce is performed using the mapper() function
- C. Placing data close to NameNode and computing it over DataNodes
- D. Compute data by placing it as close as possible



#### Which of the following defines the data locality feature of MapReduce?

- A. Reduce task in MapReduce is performed using the map() function
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- C. Placing data close to NameNode and computing it over DataNodes
- D. Compute data by placing as close as possible



The correct answer is **D** 

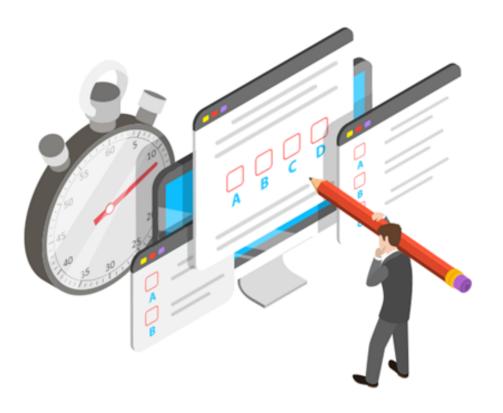
Data locality is the process of moving the computation close to where the actual data resides on the node instead of moving large data sets to the computation.



2

# What is the num parameter in the below line of code: conf.setNumMapTasks(int num)

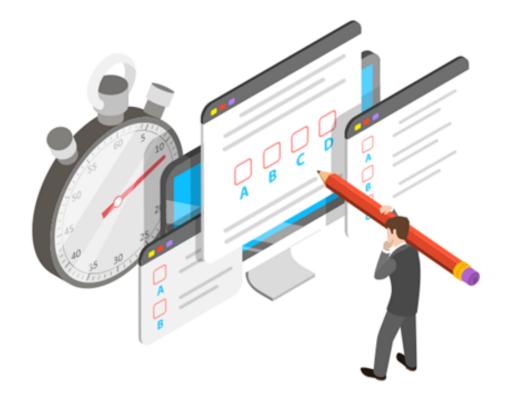
- A. Number of map tasks for this job
- B. Number of reducer tasks for this job
- C. Number of splits for this job
- D. None of the above



2

# What is the num parameter in the below line of code: conf.setNumMapTasks(int num)

- A. Number of map tasks for this job
- B. Number of reducer tasks for this job
- C. Number of splits for this job
- D. None of the above



#### The correct answer is A

The num parameter defines the number of map tasks for this job. This is only a hint at the framework. The actual number of spawned map tasks depends on the number of InputSplits generated by the job.



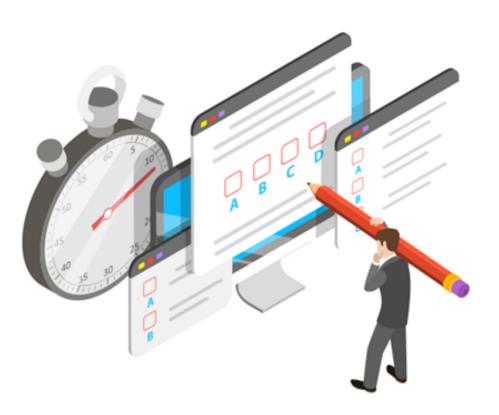
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#### Knowledge Check

# How many phases exist in MapReduce?

3

- A. ∠
- B. 5
- C. 6
- D. 2



How many phases exist in MapReduce?

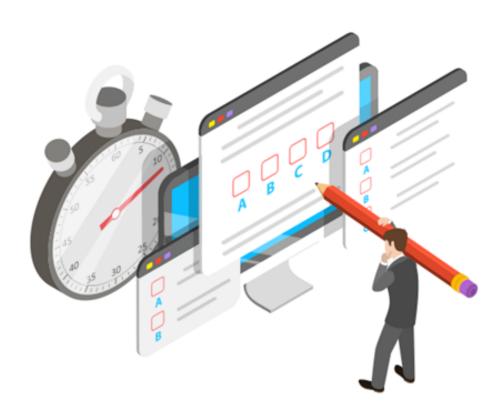
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A.

В. 5

C. 6

D.



The correct answer is **B** 



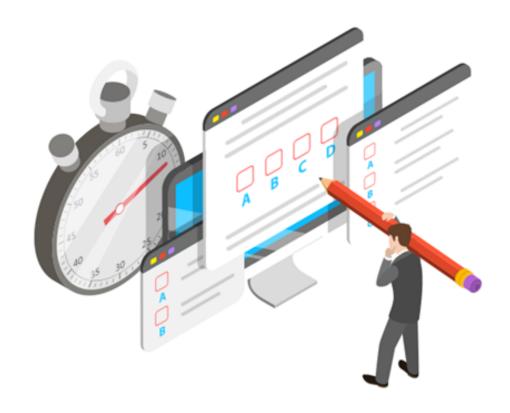
MapReduce consists of five phases: map phase, partition phase, shuffle phase, sort phase, and reduce phase.



Which of the following cannot be used as an optimization technique in MapReduce?

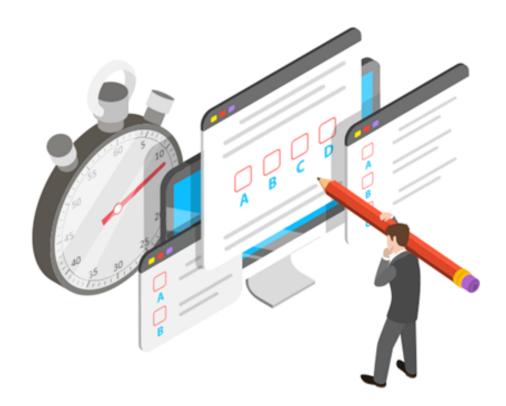
4

- A. Use of combiners
- B. LZO compression
- C. Prefer not to use writable interface for data
- D. Proper configuration of cluster



Which of the following cannot be used as an optimization technique in MapReduce?

- Use of combiners A.
- В. LZO compression
- C. Prefer not to use writable interface for data
- Proper configuration of cluster D.



The correct answer is **C** 

Prefer not to use writable interface in MapReduce is the correct option as it is not used as an optimization technique in MapReduce.



# **Lesson-End Project: Count the Number of Words Using MapReduce**

#### **Problem Scenario:**

Philip is working in a company as a Data Engineer whose focus is to count the number of times a word occurs in a corpus. He has a copy of sample data from the internet. He wants to perform a computation with a list of words and understand how many times it occurred. This can be achieved using MapReduce.

#### **Objective:**

The objective is to use MapReduce to count the number of words from a text file.

**Dataset Name**: wordcount.txt hadoop-mapreduce-example.jar



# **Lesson-End Project: Count the Number of Words Using MapReduce**

#### Tasks to Perform:

- 1. Create a sample file with the name mapreducedemo.java
- 2. Create two functions Mapper and Reduce with the main method
- Now, convert JAVA files into JAR files and give it the name "hadoop-mapreduce-example.jar"
- 4. After this, upload the JAR file into FTP
- Upload the "wordcount.txt" file into HDFS and ensure to copy the path where it has been uploaded
- 6. Run the MapReduce code in Webconsole





**Thank You**