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# **Exercise1: Setting up Environment**

**Step1**: Installing VMware Player from Training Bundle.

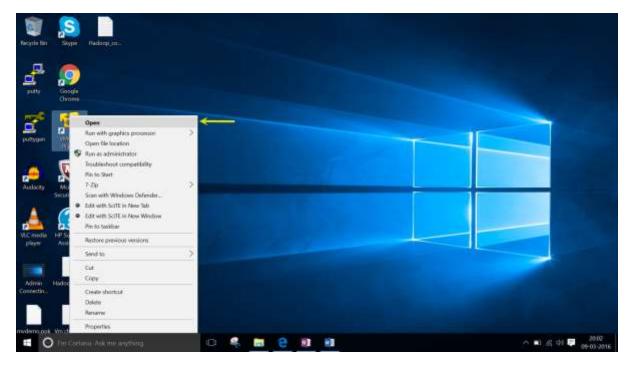
- ➤ Under 'Training\_Bundle\_For\_Hortonworks\_Developer\_Track' folder, navigate to "Additional Software" folder.
- ➤ Find the executable file named 'VmWarePlayer.exe' and complete the installation.

**Step2**: Extracting the VMware Image

- > From the same training bundle, locate the folder named VM image> 'technocrafty-quickstart-vm-5.5.0-0-vmware'
- > Unzip/Extract the files and using the VMware player browse to this location to get start with virtual machine.

Step3: Loading the VMware Image

➤ Right click and select Open or Double click on 'VMware Player' and launch the same.

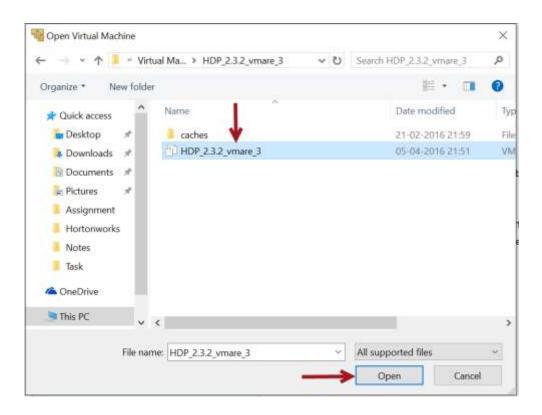


> Select the option as "Open a Virtual Machine"

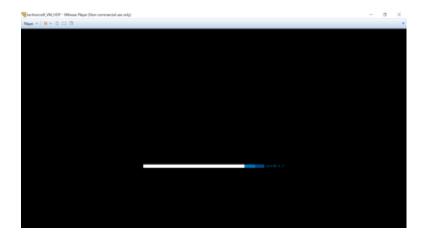


➤ Navigate to the location where you have extracted the VMware Image in earlier step.





> This will start the Virtual Machine



 $Login\ with\ user\ `cloudera'\ with\ password\ `cloudera'$ 

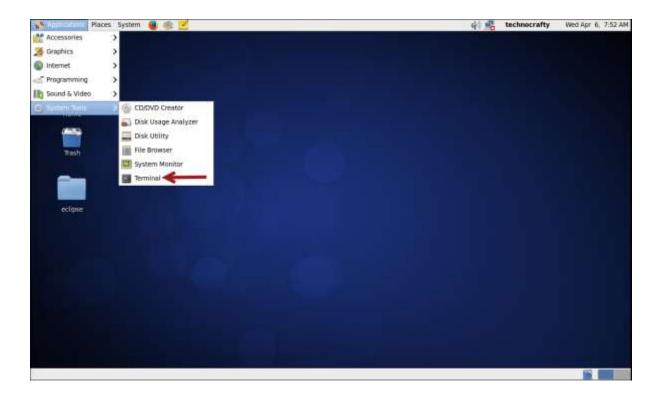
# **Exercise 2: Practicing HDFS Commands**

We will practise HDFS command line interface and web-based Hue File Browser to perform operations on files.

## **HDFS** Command line Interface

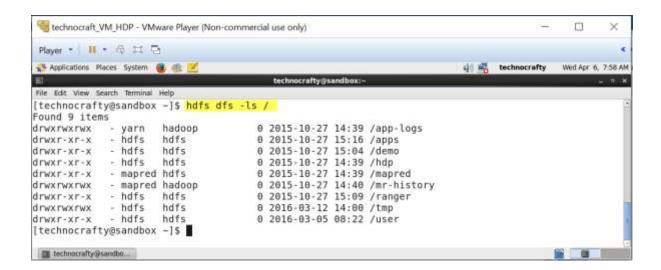
> Open terminal

Navigate to Application > System Tools > Terminal

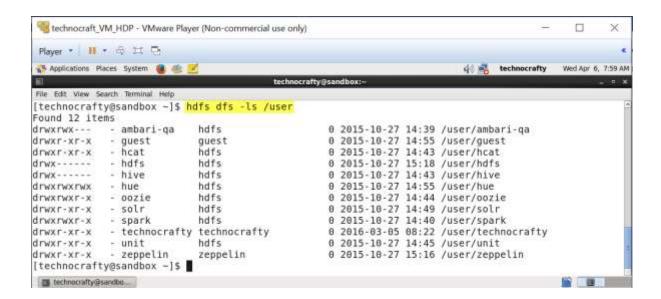


Operations on Files and Directories:

To view the content of root directory in HDFS



#### To view the content of /user directory

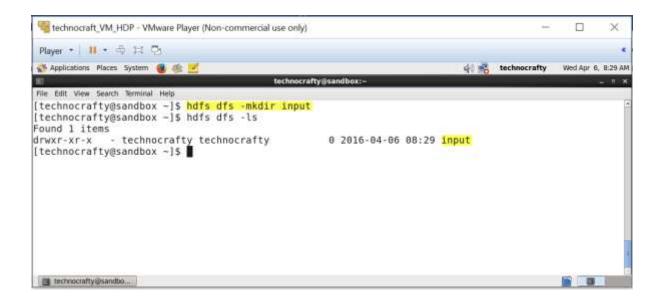




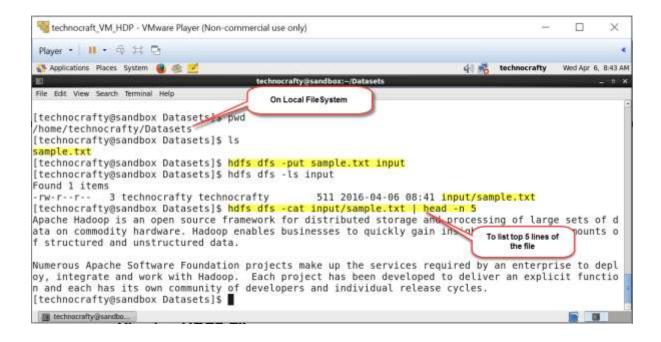
For an empty directory the prompt does not show any error while querying whereas if the directory doesn't exist it will throw an error.

#### Uploading Files to HDFS

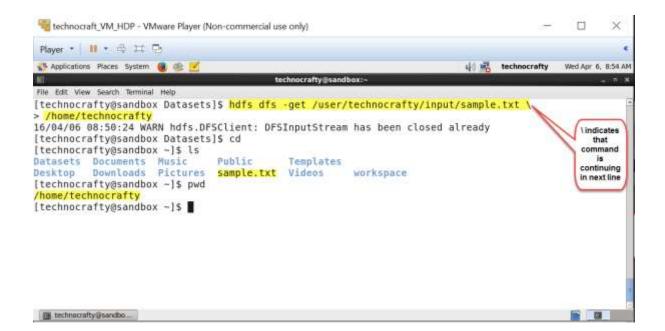
- Create a directory named 'input' in HDFS, we will use this directory throughout the exercises.
- Ensure that directory is created by running ls command



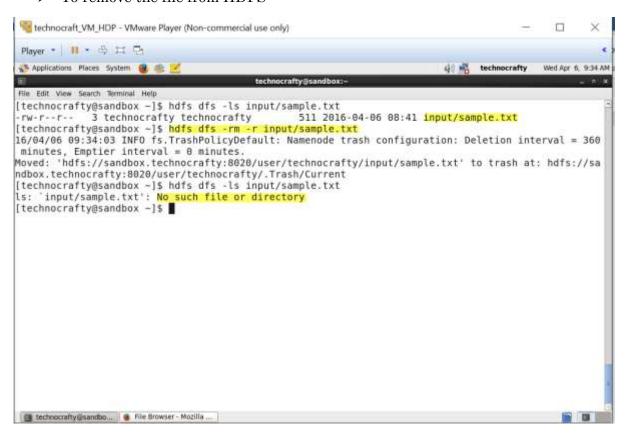
➤ Locate test file named 'sample.txt' under Datasets directory on local filesystem and upload into HDFS



> To download the file from HDFS to local filesystem



> To remove the file from HDFS



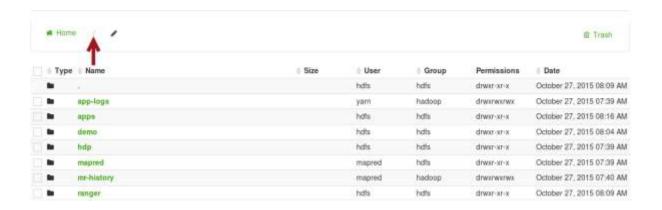
#### Web-Based Interface (Hue File Browser)

> Open a web browser on your VM, and navigate to bookmark tab and select HUE. This can be alternatively launched by specifying

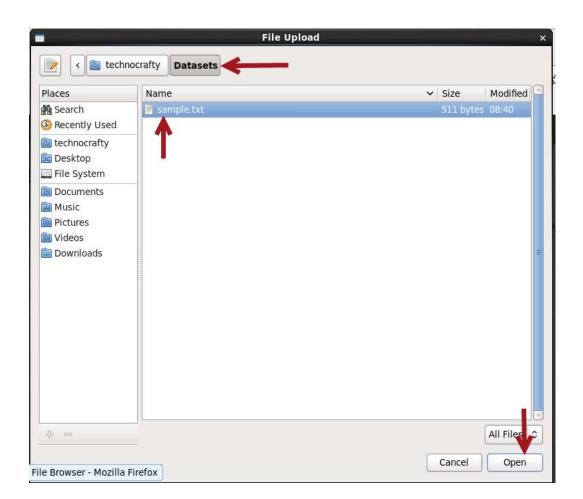
- Enter username as 'cloudera' and password 'cloudera'.
- > Navigate to file browser

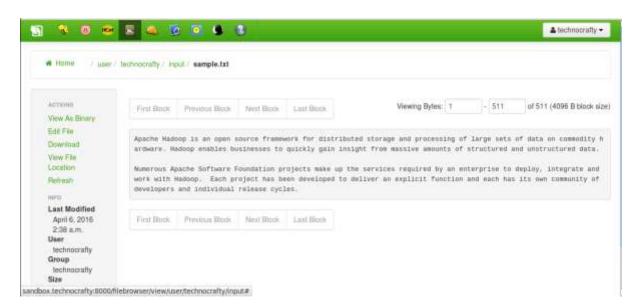
By default, the content of your HDFS home directory will be displayed. Locate the input directory created in previous step on the file browser.

Click on leading slash to view the content of root directory

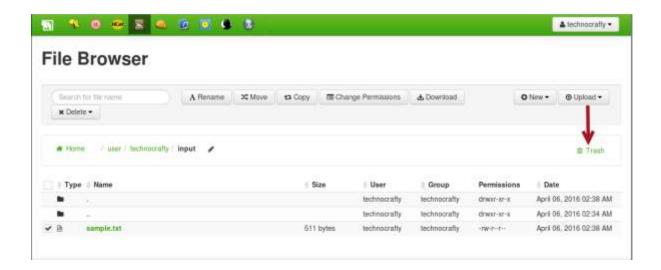


- > To perform operation on a file, select the file and several options will be enabled
- > To upload the file, click on Upload button. From the drop down you can choose to upload a plain file or zipped file
- Click on Upload > Files > Select files





> To remove the file, move it to Trash



# Exercise 3: Spark

#### Overview

In this lab, we will look at several transformations on RDD

#### Builds on

Previous labs for the transformations we'll use.

#### Run time

20-30 minutes

- > Launch spark-shell
- > Spark shell can be launched in two ways i.e. Scala and Python.
- > To launch Scala spark shell, follow below steps

```
$ cd <path>/spark-2.4.4-bin-hadoop2.6
```

\$ bin/spark-shell

> Spark creates a SparkContext object called sc, verify that the object exists

```
scala> sc
res0: org.apache.spark.SparkContext = org.apache.spark.SparkContext@8e75504
```

> To know various SparkContext methods which are available, type sc. (sc followed by dot) and then TAB key

scala> sc. accumulable addFile appName asInstanceOf broadcast clearCallSite clearJobGroup defaultParallelism files getConf getLocalProperty getRDDStorageInfo hadoopFile isInstanceOf killExecutor master newAPIHadoopRDD range runJob

setCheckpointDir

setLocalProperty

wholeTextFiles

startTime

submitJob toString

accumulableCollection applicationAttemptId binaryFiles cancelAllJobs clearFiles defaultMinPartitions emptyRDD getAllPools getExecutorMemoryStatus getPersistentRDDs getSchedulingMode hadoopRDD isLocal killExecutors metricsSystem objectFile requestExecutors sequenceFile setJobDescription setLogLevel statusTracker

tachyonFolderName

union

applicationId binaryRecords cancelJobGroup clearJars defaultMinSplits externalBlockStoreFolderName getCheckpointDir getExecutorStorageStatus getPoolForName hadoopConfiguration initLocalProperties jars makeRDD newAPIHadoopFile parallelize runApproximateJob setCallSite setJobGroup sparkUser stop textFile version

accumulator

addSparkListener

#### 1. Basic Commands and Operations

Spark is based on the concept of Resilient Distributed Dataset (RDD), which is fault tolerant collection of elements that can be operated in parallel.

#### Two ways to create RDDs:

- > Parallelizing an existing collection
- ➤ Referencing a dataset from an External Storage System

#### Parallelized collection:

To create a parallelized collection holding numbers 1 to 6

```
Example 1
val data = Array (1,2,3,4,5,6)
val distData = sc.parallelize(data)
```

```
scala> val data = Array (1,2,3,4,5,6)
data: Array[Int] = Array(1, 2, 3, 4, 5, 6)
scala> val distData = sc.parallelize(data)
distData: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[0] at parallelize at <console>:17
```

Once 'distData' dataset is created, we can perform operations such as:

Finding the sum, mean & variance

#### distData.sum

distData.mean

distData.variance

```
| Scales | distOata.sum | |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Got job 1 (sum at <console>:28 |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Got job 1 (sum at <console>:28) with 4 output partitions (allowLocal=false) |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Final stage: ResultStage 1(sum at <console>:28) |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Missing parents: List() |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Missing parents: List() |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Missing parents: List() |
| 16/04/16 | 15:08:23 | INFO MemoryStore: ensureFreeSpace(2336) called with curNem=3788, maxMem=278302556 |
| 16/04/16 | 15:08:23 | INFO MemoryStore: Block broadcast 1 stored as values in memory (estimated size 2.3 KB, free 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO MemoryStore: ensureFreeSpace(2336) called with curMem=0124, maxMem=278302556 |
| 16/04/16 | 15:08:23 | INFO MemoryStore: ensureFreeSpace(1452) called with curMem=0124, maxMem=278302556 |
| 16/04/16 | 15:08:23 | INFO MemoryStore: Block broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO MemoryStore: Block broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO SparkContext: Created broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free: 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO SparkContext: Created broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free: 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO SparkContext: Created broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free: 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO SparkContext: Created broadcast 1 piece8 stored as bytes in memory (estimated size 1452.8 B, free: 265.4 MB) |
| 16/04/16 | 15:08:23 | INFO DAGScheduler: Submitting 4 missing tasks from ResultStage 1 (MapPartitionsR00[2] at numericRDDTaDoubleRDD |
| 16/04/16 | 15:08:23 | INFO TaskSctManager: Starting task 4.0 in stage 1.0 (TID 4) | Localhost, PROCE
```

```
Scalar distDate meam

16/84/16 15:89:57 INFO SparkContext: Starting job: meam at <consoler:28

16/84/16 15:89:57 INFO SparkContext: Starting job: meam at <consoler:28

16/84/16 15:89:57 INFO SparkContext: Farents of final stage: ResultStage 2(meam at <consoler:28)

16/84/16 15:89:57 INFO Shockmeduler: Parents of final stage: List()

16/84/16 15:89:57 INFO Shockmeduler: Parents of final stage: List()

16/84/16 15:89:57 INFO Shockmeduler: Missing parents in the stage: List()

16/84/16 15:89:57 INFO Shockmeduler: Missing parents in the stage: ResultStage 2(meam at <consoler:28)

16/84/16 15:89:57 INFO Shockmeduler: Missing parents in the stage: ResultStage 2(meam at <consoler:28)

16/84/16 15:89:57 INFO MemoryStore: Block broadcast 2 stored as values in memory (estimated size 2.4 MB, free 265.4 MB)

16/84/16 15:89:57 INFO MemoryStore: Block broadcast 2 stored as values in memory (estimated size 2.4 MB, free 265.4 MB)

16/84/16 15:89:57 INFO MemoryStore: Block broadcast 2 piece8 stored as bytes in memory (estimated size 1526.8 B, free 265.4 MB)

16/84/16 15:89:57 INFO BlockManagerInfo: Added broadcast 2 piece8 stored as bytes in memory (estimated size 1526.8 B, free 265.4 MB)

16/84/16 15:89:57 INFO SparkContext: Created broadcast 2 piece8 stored as bytes in memory (estimated size 1526.8 B, free 265.4 MB)

16/84/16 15:89:57 INFO DASSCHOULER: Submitting 4 missing tasks from ResultStage 2 (MapPartitionsRDD[4] at meam at <consoler:28)

16/84/16 15:89:57 INFO TaskSetHanager: Starting task set 2.0 with 4 tasks

16/84/16 15:89:57 INFO TaskSetHanager: Starting task 8.0 in stage 2.0 (TID 9, localhost, PROCESS LOCAL, 1313 bytes)

16/84/16 15:89:57 INFO TaskSetHanager: Starting task 8.0 in stage 2.0 (TID 9, localhost, PROCESS LOCAL, 1313 bytes)

16/84/16 15:89:57 INFO TaskSetHanager: Starting task 9.0 in stage 2.0 (TID 8)

16/84/16 15:89:57 INFO TaskSetHanager: Starting task 9.0 in stage 2.0 (TID 8)

16/84/16 15:89:57 INFO TaskSetHanager: Finished task 8.0 in stage 2.0 (TID 8)

16/84/16 15:89:57 INFO TaskSetHanager: Finished
```

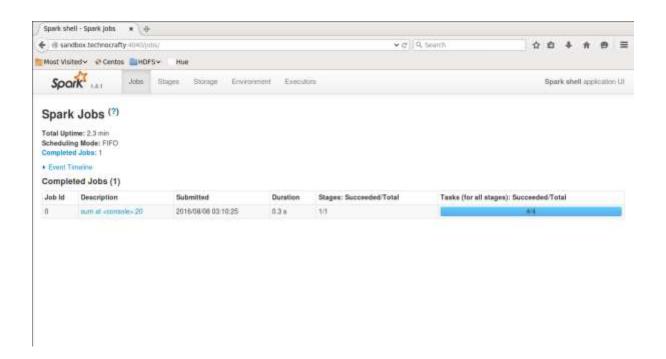
```
distData.variance
16/04/16 15:11:01 INFO DAGScheduler: Got job 3 (variance at <console>:20
16/04/16 15:11:01 INFO DAGScheduler: Final stage: ResultStage 3(variance at <console>:20) with 4 output partitions (allowLocal=false)
16/04/16 15:11:01 INFO DAGScheduler: Parents of final stage: List()
16/04/16 15:11:01 INFO DAGScheduler: Parents of final stage: List()
16/04/16 15:11:01 INFO DAGScheduler: Missing parents: List()
16/04/16 15:11:01 INFO DAGScheduler: Submitting ResultStage 3 (MapPartitionsRDD[6] at variance at <console>:20), which has no n
16/04/16 15:11:01 INFO DAGScheduler: Submitting ResultStage 3 (MapPartitionsRDD[6] at variance at <console>:20), which has no n
16/04/16 15:11:01 INFO MemoryStore: ensureFreeSpace(2496) called with curMem=11598, maxMem=278302556
16/04/16 15:11:01 INFO MemoryStore: ensureFreeSpace(1252) called with curMem=14094, maxMem=278302556
16/04/16 15:11:01 INFO MemoryStore: ensureFreeSpace(1252) called with curMem=14094, maxMem=278302556
16/04/16 15:11:01 INFO MemoryStore: ensureFreeSpace(1252) called with curMem=14094, maxMem=278302556
16/04/16 15:11:01 INFO MemoryStore: ensureFreeSpace(1252) called with curMem=14094, maxMem=278302556
16/04/16 15:11:01 INFO BlockManagerInfo: Added broadcast 3 piece0 stored as bytes in memory (estimated size 1529.0 8, free 265.4 MB)
16/04/16 15:11:01 INFO BlockManagerInfo: Added broadcast 3 piece0 in memory on localhost:58083 (size: 1529.0 8, free: 265.4 MB)
16/04/16 15:11:01 INFO BlockManagerInfo: Added broadcast 3 piece0 in memory on localhost:58083 (size: 1529.0 8, free: 265.4 MB)
16/04/16 15:11:01 INFO TaskSctManager: Storting task store as with 4 tasks
16/04/16 15:11:01 INFO TaskSctManager: Storting task 8ct 3.0 with 4 tasks
16/04/16 15:11:01 INFO TaskSctManager: Storting task 8ct 3.0 with 4 tasks
16/04/16 15:11:01 INFO TaskSctManager: Storting task 8ct 3.0 with 4 tasks
16/04/16 15:11:01 INFO TaskSctManager: Storting task 8ct 3.0 with 4 tasks
16/04/16 15:11:01 INFO TaskSctManager: Storting task 8ct 3.0 with 4 tasks
16/04/16 15:11:01 INFO TaskSc
```



Spark sets the number of partition (i.e.to cut the dataset into) automatically based on your cluster, but this can be set manually by passing second parameter to *parallelize* (e.g. sc.parallelize(data,10))

## Spark UI

Spark UI can be accessed on port <localhost:4040> Job is created once you perform an action.



#### External Datasets:

Spark can create distributed datasets from any storage system supported by Hadoop, Spark supports text files, SequenceFiles and any other Hadoop InputFormat.

To load a file from your Local Filesystem

# $Example 2 \\ val \ textFile = sc.textFile ("<\underline{path}>/\underline{Datasets/Employee.txt}")$



While using local filesystem, the file must be accessible at the same path on worker nodes.

Operations on RDD will be discussed in next section.



Spark's file-based input methods supports directories, compressed files and wildcards as well. i.e. textFile("/my/directory"), textFile("/my/directory/\*.txt"), and textFile("/my/directory/\*.gz")

## 2. Operations on RDD

RDDs supports two types of operations:

#### > Transformations

Creates a new dataset from an existing one

#### > Actions

Returns a value to the driver program after running a computation on the dataset.

Example: map is a transformation that passes each dataset element through a function and return a new RDD, whereas reduce is an action that aggregates all the elements of RDD using function and returns final result.



All transformations in Spark are lazy i.e. transformations are only computed when an action requires a result to be returned.

textFile dataset was already created in previous step, lets perform below operations

> counting the occurrence of lines having a particular word in it

Example: filter using scala shell

val textFile = sc.textFile("<path>/Datasets/Employee.txt")

textFile.count() //number of items in this RDD

textFile.first() //First item in this RDD

val linesWithHadoop = textFile.filter(line => line.contains("7839"))

linesWithHadoop.count()

> Add up the sizes of all the lines

```
val lineLength = textFile.map(s => s.length)
val totalLength = lineLength.reduce((a, b) => a + b)
```

> Line with maximum words

# Example:

textFile.map(line => line.split(" ").size).reduce((a,b) => if(a>b) a else b)

```
Example4: filter

val sampleData = 1 to 5000

val totData = sc.parallelize(sampleData)

val result = totData.filter(_ %2==0)

result.collect()

with number of partition = 2

val totDataPar = sc.parallelize(sampleData,2)

val resultPar = totDataPar.filter(_ %2==0)

resultPar.collect()
```

```
scala> val totData =sc.parallelize(sampleData)
totData: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDO[I2] at parallelize at <console>:17
scala> val result = totData.filter( %2=0)
16/04/16 15:41:40 INFO BlockManagerInfo: Removed broadcast 9 piece0 on localhost:58083 in memory (size: 1983.8 B, free: 265.4 MB)
result: org.apache.spark.rdd.RDO[Int] = MapPartitionsRDD[I3] at filter at <console>:19
```

```
Incomplete Sealt.collect()

15/84/16 15-42:42 18F0 SparkContext: Starting job: collect at *console*:22

15/84/16 15-42:42 18F0 DASScheduler: sof job 9 (cellect at *console*:22) with 4 output partitions (allowLocalefalse)

15/84/16 15-42:42 18F0 DASScheduler: Parents of final stage List()

15/84/16 15-42:42 18F0 DASScheduler: Parents of final stage List()

15/84/16 15-42:42 18F0 DASScheduler: Submitting ResultStage 9 (MapPartitionsRDD[15] at filter at *console*:191, which has no missing parents

15/84/16 15-42:42 18F0 DASScheduler: Submitting ResultStage 9 (MapPartitionsRDD[15] at filter at *console*:191, which has no missing parents

15/84/16 15-42:42 18F0 MemoryStore: Block broadcast 10 stored as *values in sembry (estimated size 187.6.8.8. free 265.2 MB)

15/84/16 15-42:42 18F0 MemoryStore: Block broadcast 10 stored as *values in sembry (estimated size 187.6.8.8. free 265.2 MB)

15/84/16 15-42:42 18F0 MemoryStore: Block broadcast 10 stored as *values in sembry (estimated size 187.6.8.8. free 265.2 MB)

15/84/16 15-42:42 18F0 MemoryStore: Block broadcast 10 plexed stored as bytes in memory (estimated size 187.6.8.8. free 265.2 MB)

15/84/16 15-42:42 18F0 MemoryStore: Slock broadcast 10 plexed stored as bytes in memory (estimated size 187.6.8. free 265.2 MB)

15/84/16 15-42:42 18F0 DASScheduler: Simbitting 4 missing tasks from ResultStage 9 (MapPartItionsRDD[13] at filter at *console*:19)

15/84/16 15-42:42 18F0 DASScheduler: Simbitting 4 missing tasks from ResultStage 9 (MapPartItionsRDD[13] at filter at *console*:19)

15/84/16 15-42:42 18F0 DASScheduler: Simbitting 4 missing tasks 9.8 in stage 9.8 (TID 25) localbost, PROCESS LOCAL, 1399 bytes)

16/84/16 15-42:42 18F0 DASScheduler: Simbitting 4 missing tasks 9.8 in stage 9.8 (TID 25) localbost, PROCESS LOCAL, 1399 bytes)

16/84/16 15-42:42 18F0 DASScheduler: Simbitting 4 missing tasks 9.8 in stage 9.8 (TID 25) localbost, PROCESS LOCAL, 1426 bytes)

16/84/16 15-42:42 18F0 DASScheduler: MapPartiting 4 missing process 19 missing 9.8 (TID 25) localbost, PROCESS LO
```

```
Example: Scala example of map

val input = sc.parallelize(List(1, 2, 3, 4))

val result = input.map(x => x * x)

println(result.collect().mkString(","))
```

```
scala> val input = sc.parallelize(List(I, 2, 3, 4))
input: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[16] at parallelize at <console>:15
scala> val result = input.map(x => x * x)
result: org.apache.spark.rdd.RDD[Int] = MapPartitionsRDD[17] at map at <console>:17
```

```
Scala> println(result.collect().mkString(*,*))
16/04/16 15:47:45 INFO SparKcontext: Starting job: collect at <console>:20
16/04/16 15:47:45 INFO DAGScheduler: Got job 11 {collect at <console>:20 with 4 output partitions (allowLocal=false)
16/04/16 15:47:45 INFO DAGScheduler: Final stage: ResultStage 11(collect at <console>:20)
16/04/16 15:47:45 INFO DAGScheduler: Parents of final stage: List()
16/04/16 15:47:45 INFO DAGScheduler: Missing parents: List()
16/04/16 15:47:45 INFO DAGScheduler: Submitting ResultStage 11 (MapPartitionsRDD[17] at map at <console>:17), which has
16/04/16 15:47:45 INFO McMoryStore: Block broadcast 12 stored as values in memory (estimated size 1944.0 B, free 265.2
16/04/16 15:47:45 INFO MemoryStore: ensureFreeSpace[1944] called with curMem=221505, maxMem=278302556
16/04/16 15:47:45 INFO MemoryStore: ensureFreeSpace[2122) called with curMem=223509, maxMem=278302556
16/04/16 15:47:45 INFO MemoryStore: ensureFreeSpace[2122) called with curMem=223509, maxMem=278302556
16/04/16 15:47:45 INFO MemoryStore: ensureFreeSpace[2122] called with curMem=223509, maxMem=278302556
16/04/16 15:47:45 INFO BlockManagerInfo: Added broadcast 12 piece0 stored as bytes in memory (estimated size 1212.0 B, free 16/04/16 15:47:45 INFO BlockManagerInfo: Added broadcast 12 piece0 in memory on localhost:58883 (size: 1212.0 B, free 16/04/16 15:47:45 INFO BlockManagerInfo: Added broadcast 12 piece0 in memory on localhost:58883 (size: 1212.0 B, free 16/04/16 15:47:45 INFO DAGScheduler: Submitting 4 missing tasks from ResultStage 11 (MapPartitionsRDD[17] at map at <coll>
16/04/16 15:47:45 INFO TaskSchedmanger: Starting task set 11.0 with 4 tasks
16/04/16 15:47:45 INFO TaskSchedmanager: Starting task 1.0 in stage 11.0 (TID 32, localhost, PROCESS_LOCAL, 1313 bytes)
16/04/16 15:47:45 INFO TaskSchedmanager: Starting task 1.0 in stage 11.0 (TID 33, localhost, PROCESS_LOCAL, 1313 bytes)
16/04/16 15:47:45 INFO TaskSchedmanager: Starting task 1.0 in stage 11.0 (TID 33, localhost, PROCESS_LOCAL, 1313 bytes)
16/04/16 15:47:45 INFO
```

```
Example 6: union

val seta = sc.parallelize(1 to 10)

val setb = sc.parallelize(5 to 15)

(seta union setb).collect
```

```
Scala> val seta = Sc.parallelize(1 to 10)

16/84/16 15:58:87 INFO BlockManagerInfo: Removed broadcast 12 piece0 on localhost:58883 in memory (size: 1212.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO BlockManagerInfo: Removed broadcast 11 piece0 on localhost:58883 in memory (size: 1209.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO BlockManagerInfo: Removed broadcast 11 piece0 on localhost:58883 in memory (size: 1209.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO BlockManagerInfo: Removed broadcast 11 piece0 on localhost:58883 in memory (size: 1212.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO BlockManagerInfo: Removed broadcast 11 piece0 on localhost:58883 in memory (size: 1212.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO BlockManagerInfo: Removed broadcast 12 piece0 on localhost:58883 in memory (size: 1212.0 B, free: 265.4 MB)

16/84/16 15:58:07 INFO TaskSetManager: Finished task 6.0 in stage 12.0 (TID 41) in 17 ms on localhost (6/8)

16/84/16 15:50:40 INFO TaskSetManager: Finished task 7.0 in stage 12.0 (TID 42) in 10 ms on localhost (7/8)

16/84/16 15:50:40 INFO TaskSetManager: Finished task 7.0 in stage 12.0 (TID 42) in 10 ms on localhost (7/8)
```

16/04/16 15:50:40 INFO TaskSetManager: Finished task 7.0 in stage 12.0 (TID 42) in 10 in 50 in totalhost (7/0) 16/04/16 15:50:40 INFO TaskSetManager: Finished task 5.0 in stage 12.0 (TID 40) in 27 ms on localhost (8/0) 16/04/16 15:50:40 INFO TaskSchedulerImpl: Removed TaskSet 12.0, whose tasks have all completed, from pool 16/04/16 15:50:40 INFO DAGScheduler: ResultStage 12 (collect at <console>:20) finished in 0.047 s 16/04/16 15:50:40 INFO DAGScheduler: Job 12 finished: collect at <console>:20, took 0.071821 s res12: Array[Int] = Array(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15)

```
Example 7: flatMap() using scala shell

Splitting lines into multiple words

val lines = sc.parallelize(List("hello world", "hi"))

val words = lines.flatMap(line => line.split(" "))

words.first() // returns "hello"
```

```
scala> val lines = sc.parallelize(List("hello world", "hi"))
lines: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[21] at parallelize at <console>:15
scala> val words = lines.flatMap(line => line.split(" "))
words: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[22] at flatMap at <console>:17
```

```
scala> words.first()
16/04/16 15:54:16 INFO SparkContext: Starting job: first at <console>:20
16/04/16 15:54:16 INFO DAGScheduler: Got job 13 (first at <console>:20) with 1 output partitions (allowLocal=true)
16/04/16 15:54:16 INFO DAGScheduler: Final stage: ResultStage 13(first at <console>:20)

16/04/16 15:54:16 INFO TaskSetManager: Finished task 2.0 in stage 14.0 (TID 46) in 22 ms on localhost (2/3)
16/04/16 15:54:16 INFO TaskSetManager: Finished task 1.0 in stage 14.0 (TID 45) in 25 ms on localhost (3/3)
16/04/16 15:54:16 INFO TaskSchedulerImpl: Removed TaskSet 14.0, whose tasks have all completed, from pool
16/04/16 15:54:16 INFO DAGScheduler: ResultStage 14 (first at <console>:20) finished in 0.021 s
16/04/16 15:54:16 INFO DAGScheduler: Job 14 finished: first at <console>:20, took 8.063394 s
res13: String = hello
```

#### 3. Paired RDD

- > Spark provides special operations on RDDs containing key/value pairs. These RDDs are called pair RDD.
- ➤ Pair RDD allows you to act on each key in parallel or regroup data across the network.

```
Example 8: foldByKey

val a = sc.parallelize(List("kim","kumar","muthu","tim","lak","vamsi"),2)

val b = a.map(x => (x.length,x))

b.collect

b.foldByKey("")(_+_).collect
```

```
scala> val a = sc.parallelize(List("kim","kumar","muthu","tim","lak","vamsi"),2)
a: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[24] at parallelize at <console>:15
scala> val b = a.map(x => (x.length,x))
b: org.apache.spark.rdd.RDD[(Int, String)] = MapPartitionsRDD[25] at map at <console>:17
```

```
Scala> b. collect

16/84/16 16:04:40 INFO SparkContext: Starting jub: collect at <console>:20

15/84/16 16:04:40 INFO DAGScheduler: Got jub 15 (collect at <console>:20) with 2 output partitions (allowLocal=false)

16/84/16 16:04:40 INFO DAGScheduler: Final stage: ResultStage 15(collect at <console>:20)

16/84/16 16:04:40 INFO DAGScheduler: Missing parents: List()

16/84/16 16:04:41 INFO DAGScheduler: Missing parents: List()

16/84/16 16:04:41 INFO DAGScheduler: Submitting ResultStage 15 (MapPartitionsRDD[25] at map at <console>:17), which has no missing

16/84/16 16:04:41 INFO MemoryStore: ensureFreeSpace(1912) called with curMem=218412, maxMem=278302556

16/84/16 16:04:41 INFO MemoryStore: ensureFreeSpace(1912) called with curMem=220324, maxMem=278302556

16/84/16 16:04:48 INFO MemoryStore: ensureFreeSpace(1185) called with curMem=220324, maxMem=278302556

16/84/16 16:04:48 INFO MemoryStore: Block broadcast 16 stored as bytes in memory (estimated size 1185.6 B, free 265.2 MB)

16/84/16 16:04:48 INFO MemoryStore: Block broadcast 16 piece® stored as bytes in memory (estimated size 1185.6 B, free 265.2 MB)

16/84/16 16:04:48 INFO SparkContext: Created broadcast 16 piece® in memory on localhost:50003 (size: 1185.0 B, free: 265.4 MB)

16/84/16 16:04:48 INFO SparkContext: Created broadcast 16 from broadcast at DAGScheduler. scale:874

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task set 15.0 with 2 tasks

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task set 15.0 with 2 tasks

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task set 15.0 with 2 tasks

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task from Broadcast at DAGScheduler. scale:874

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task from Broadcast 15.0 (TID 48)

16/84/16 16:04:48 INFO TaskSchedulerImpl: Adding task 18 in stage 15.0 (TID 48)

16/84/16 16:04:48 INFO TaskSchedulerImpl task 1.0 in stage 15.0 (TID 48)

16/84/16 16:04:48 INFO TaskSchedulerImpl task 1.0 in stage 15.0 (TID 48)

16/84/16 16:04:48 INFO TaskSchedulerImpl task 1.0 in
```

```
scala> b.foldByKey(")( + ).collect
16/04/16 16:05:45 INFO SparkContext: Starting job: collect at <console>:20
16/04/16 16:05:45 INFO DAGScheduler: Registering RDD 25 (map at <console>:17)
16/04/16 16:05:45 INFO DAGScheduler: Got job 16 (collect at <console>:20) with 2 output partitions (allowLocal=false)
16/04/16 16:05:45 INFO DAGScheduler: Final stage: ResultStage 17(collect at <console>:20)

16/04/16 16:05:45 INFO Executor: Finished task 0.0 in stage 17.0 (TID 51). 882 bytes result sent to driver
16/04/16 16:05:45 INFO Executor: Finished task 1.0 in stage 17.0 (TID 52). 1070 bytes result sent to driver
16/04/16 16:05:45 INFO TaskSetManager: Finished task 0.0 in stage 17.0 (TID 51) in 60 ms on localhost (1/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 59 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 50 ms on localhost (2/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 50 ms on localhost (1/2)
16/04/16 16:05:45 INFO TaskSetManager: Finished task 1.0 in stage 17.0 (TID 52) in 50 ms on localhost (1/2)
16/04/16 16:05:45 INFO TaskSet
```

```
Example 9: foldByKey
val deptEmployees =
 List
 ("dept1",("kumar1",1000.0)),
 ("dept1",("kumar2",1200.0)),
 ("dept2",("kumar3",2200.0)),
 ("dept2",("kumar4",1400.0)),
 ("dept2",("kumar5",1000.0)),
 ("dept2",("kumar6",800.0)),
 ("dept1",("kumar7",2000.0)),
 ("dept1",("kumar8",1000.0)),
 ("dept1",("kumar9",500.0))
val employeeRDD = sc.makeRDD(deptEmployees)
val maxByDept = employeeRDD.foldByKey(("dummy",Double.MinValue))((acc,element)
=> if(acc._2 > element._2)acc else element)
println("Maximum salaries in each dept" + maxByDept.collect().toList)
```

```
scala> val maxByDept = employeeRDD.foldByKey({*dummy*,Double.MinValue)}{(acc,element)=>if(acc, 2 > element, 2)acc else element)
maxByDept: org.apache.spark.rdd.RDD[{String, {String, Double}}] = ShuffledRDD[29] at foldByKey at <console>:19
```

```
scala> println("Maximum salaries in eachdept" + maxByDept.collect().toList)
16/04/16 16:13:51 INFO SparkContext: Starting job: collect at <console>:22
16/04/16 16:13:51 INFO DAGScheduler: Registering RDD 27 (makeRDD at <console>:17)
16/04/16 16:13:51 INFO DAGScheduler: Got job 17 (collect at <console>:22) with 4 output partitions (allowLocal=false)
16/04/16 16:13:51 INFO DAGScheduler: Final stage: ResultStage 19(collect at <console>:22)
16/04/16 16:13:51 INFO DAGScheduler: Parents of final stage: List(ShuffleMapStage 18)
16/04/16 16:13:51 INFO DAGScheduler: Missing parents: List(ShuffleMapStage 18)
```

```
16/04/16 16:13:51 INFO Executor: Finished task 3.0 in stage 19.0 (TID 60). 882 bytes result sent to driver 16/04/16 16:13:51 INFO TaskSetManager: Finished task 3.0 in stage 19.0 (TID 60) in 12 ms on localhost (3/4) 16/04/16 16:13:51 INFO ShuffleBlockFetcherIterator: Getting 2 non-empty blocks out of 4 blocks 16/04/16 16:13:51 INFO ShuffleBlockFetcherIterator: Started 0 remote fetches in 0 ms 16/04/16 16:13:51 INFO Executor: Finished task 0.0 in stage 19.0 (TID 57). 1050 bytes result sent to driver 16/04/16 16:13:51 INFO TaskSetManager: Finished task 0.0 in stage 19.0 (TID 57) in 35 ms on localhost (4/4) 16/04/16 16:13:51 INFO TaskSetManager: Finished task 0.0 in stage 19.0 (TID 57) in 35 ms on localhost (4/4) 16/04/16 16:13:51 INFO DAGSchedulerImpl: Removed TaskSet 19.0, whose tasks have all completed, from pool 16/04/16 16:13:51 INFO DAGScheduler: ResultStage 19 (collect at <console>:22) finished in 0.035 s 16/04/16 16:13:51 INFO DAGScheduler: Job 17 finished: collect at <console>:22, took 0.086547 s Haximum salaries in eachdeptList((dept1,(kumar7,2000.0)), (dept2,(kumar3,2200.0)))
```

```
Example 10: reduceByKey

val textFile = sc.textFile("file:/home/technocrafty/Datasets/sample.txt ")

val counts = textFile.flatMap(line => line.split(" "))

counts.collect

val counts = textFile.flatMap(line => line.split(" ")).map(word => (word,1))

counts.collect

val counts = textFile.flatMap(line => line.split(" ")).map(word => (word,1)).reduceByKey(_+_)

counts.collect.foreach(println)
```

```
scala> val counts = textFile.flatMap(line => line.split(" "))
counts: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[30] at flatMap at <console>:17
```

```
in/04/16 16:17.16 DWTO SparkContext: Starting job: collect at *console*:28

16/04/16 16:17.16 DWTO SparkContext: Starting job: collect at *console*:28

16/04/16 16:17.16 DWTO SparkContext: Starting job: collect at *console*:28| with 2 output partitions (allowLocal=false)

16/04/16 16:17.16 DWTO SparkContext: final stage : Resultings : Review and *console*:28|

16/04/16 16:17.16 DWTO SparkContext: final stage : Resultings : Review and *console*:28|

16/04/16 16:17.16 DWTO SparkContext: Substiting Resulting Resultings Resulting Resu
```

```
scala> val counts = textFile.flatMap(line => line.split(* *)).map(word => (word,1))
counts: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[32] at map at <console>:17
```

```
scala* val counts = textFile.flatMap(line *> line.split(' ')).map(word -> (word,1)).reduceByKey( + )

16/84/16 16:20:40 INFO BlockManagerInfo: Removed broadcast 22 piece0 on localhost:58883 in memory [size: 1962.8 8, free: 265.4 MB)

16/84/16 16:20:40 INFO BlockManagerInfo: Removed broadcast 21 piece0 on localhost:58883 in memory [size: 1927.0 8, free: 265.4 MB)

16/84/16 16:20:40 INFO BlockManagerInfo: Removed broadcast 20 piece0 on localhost:58883 in memory [size: 2.8 KB, free: 265.4 MB)

16/84/16 16:20:40 INFO BlockManagerInfo: Removed broadcast 19 piece0 on localhost:58883 in memory [size: 1737.8 B, free: 265.4 MB)

counts: org.apache.spark.rdd.RDD[(String, Int)] = ShuffledRDD[35] at reduceByKey at «console»:17
```

```
scala> counts.collect.foreach(println)
16/04/16 16:21:24 INFO SparkContext: Starting job: collect at <console>:28
16/04/16 16:21:24 INFO DAGScheduler: Registering RDD 34 (map at <console>:17)
16/04/16 16:21:24 INFO DAGScheduler: Got job 20 (collect at <console>:20) with 2 output partitions (allowLocal=false)
16/04/16 16:21:24 INFO DAGScheduler: Final stage: ResultStage 23(collect at <console>:20)
16/04/16 16:21:24 INFO DAGScheduler: Parents of final stage: List(ShuffleMapStage 22)
```

Final print output logs are too big to display here, but the content will look as below

```
(integrate,1)
(by, 1)
(individual,1)
(storage, 1)
(structured.1)
(for, 1)
(Each, 1)
(amounts,1)
(cycles.,1)
(an, 3)
(Foundation, 1)
(distributed,1)
(and,5)
(required,1)
(deploy,,1)
(the, 1)
(Hadoop, 2)
scala>
```

#### Example 11: reduceByKey

 $val \ myRDD = sc.parallelize(Seq((1,"A"),(2,"B"),(2,"D"),(3,"C"),(3,"A"),(3,"B"),(3,"A")),1) \\ val \ resultRDD = myRDD.reduceByKey((x, y) => \{println("x"+x+"::"+"y"+y);x+y\}) \\ resultRDD.foreach(println)$ 

```
scala> val myRDD =sc.parallelize(Seq((1,"A"),(2,"B"),(2,"D"),(3,"C"),(3,"A"),(3,"B"),(3,"A")),1)
myRDD: org.apache.spark.rdd.RDD((Int, String)) = ParallelCollectionRDD(38) at parallelize at <console>:15
scala> val resultRDD = myRDD.reduceByKey((x,y) => {println("x"+x+"::"+"y"+y);x+y})
resultRDD: org.apache.spark.rdd.RDD((Int, String)) = ShuffledRDD[39] at reduceByKey at <console>:17
```

```
Scala> resultRDD.foreach(println)
16/04/16 16:25:03 INFO SparkContext: Starting job: foreach at <console>:20
16/04/16 16:25:03 INFO DAGScheduler: Registering RDD 38 (parallelize at <console>:15)
16/04/16 16:25:03 INFO DAGScheduler: Got job 21 (foreach at <console>:20) with 1 output partitions (allowLocal=false)

16/04/16 16:25:04 INFO ShuffleBlockFetcherIterator: Getting 1 non-empty blocks out of 1 blocks
16/04/16 16:25:04 INFO ShuffleBlockFetcherIterator: Started 0 remote fetches in 0 ms
(1.A)
(3.CABA)
(2.BD)
16/04/16 16:25:04 INFO Executor: Finished task 0.0 in stage 25.0 (TID 70). 806 bytes result sent to driver
16/04/16 16:25:04 INFO TaskSchedulerImpl: Removed TaskSet 25.0, whose tasks have all completed, from pool
16/04/16 16:25:04 INFO DAGScheduler: ResultStage 25 (foreach at <console>:20) finished in 0.000>:20 finished in 0.0000>:20 finished in 0.0000>:20 finished in 0.0000>:20 finished finished 0.0000>:20 finished in 0.0000>:20 finished in 0.0000>:20 finished in 0.0000>:20 finished in 0.00000>:20 finished in 0.00000>:20 finished in 0.000000>:20 finished in 0.00000000000000000
```

```
Example 12: collectAsMap
```

```
\label{eq:val_myrdd} \mbox{val myRDD} = \mbox{sc.parallelize}(\mbox{Seq}((1,"A"),(2,"B"),(2,"D"),(3,"C"),(3,"A"),(3,"B"), \\ (3."A")))
```

myRDD.collectAsMap()

#### Example 13: countByKey

#### myRDD.countByKey()

```
scala> myRDD.countByKey()
16/04/16 16:28:48 INFO SparkContext: Starting job: countByKey at <console>:18
16/04/16 16:28:48 INFO DAGSCheduler: Registering RDD 41 (countByKey at <console>:18)
16/04/16 16:28:48 INFO DAGSCheduler: Got job 23 (countByKey at <console>:18) with 4 output partitions (allowLocal=false)
16/04/16 16:28:48 INFO DAGSCheduler: Final stage: ResultStage 28(countByKey at <console>:18)

16/04/16 16:28:48 INFO TaskSetManager: Finished task 0.0 in stage 28.0 (TID 79) in 13 ms on localhost (2/4)
16/04/16 16:28:48 INFO TaskSetManager: Finished task 3.0 in stage 28.0 (TID 82) in 12 ms on localhost (3/4)
16/04/16 16:28:40 INFO TaskSetManager: Finished task 1.0 in stage 28.0 (TID 80). 1075 bytes result sent to driver
16/04/16 16:28:40 INFO TaskSetManager: Finished task 1.0 in stage 28.0 (TID 80) in 15 ms on localhost (4/4)
16/04/16 16:28:40 INFO TaskSchedulerImpl: Removed TaskSet 28.0, whose tasks have all completed, from pool
16/04/16 16:28:40 INFO DAGSCheduler: ResultStage 28 (countByKey at <console>:18, took 0.857667 s
16/04/16 16:28:40 INFO DAGSCheduler: Job 23 finished: countByKey at <console>:18, took 0.857667 s
```

#### Example 14: groupBy

```
val a = sc.parallelize(1 to 15) a.groupBy(x => \{if (x \% 2 == 0) \text{ "even" else "odd"}\}).collect
```

```
scala> val a = sc.parallelize(1 to 15)
a: org.apache.spark.rdd.RDO[Int) = ParallelCollectionRDD[8] at parallelize at <console>:15

scala> a.groupBy(x => {if {x % 2 == 0} "even" else"add"}}.collect
16/04/16 16:30:52 INFO DAGScheduler: Registering RDD 1 (groupBy at <console>:18)
16/04/16 16:30:52 INFO DAGScheduler: Registering RDD 1 (groupBy at <console>:18)
16/04/16 16:30:52 INFO DAGScheduler: Got job 8 (collect at <console>:18) with 4 output partitions (allowLocal=false)

16/04/16 16:30:53 INFO TaskSetManager: Finished task 8:0 in stage 1.0 (TID 5) in 98 ms on localhost (1/4)
16/04/16 16:30:53 INFO TaskSetManager: Finished task 1.0 in stage 1.0 (TID 5) in 98 ms on localhost (2/4)
16/04/16 16:30:53 INFO TaskSetManager: Finished task 2.0 in stage 1.0 (TID 6). 1467 bytes result sent to driver
16/04/16 16:30:53 INFO Executor: Finished task 3.0 in stage 1.0 (TID 7). 1466 bytes result sent to driver
16/04/16 16:30:53 INFO TaskSetManager: Finished task 3.0 in stage 1.0 (TID 7) in 180 ms on localbost (3/4)
16/04/16 16:30:53 INFO TaskSetManager: Finished task 3.0 in stage 1.0 (TID 7) in 111 ms on localbost (4/4)
16/04/16 16:30:53 INFO TaskSetManager: Finished task 3.0 in stage 1.0 (TID 7) in 111 ms on localbost (4/4)
16/04/16 16:30:53 INFO TaskSetManager: Finished task 3.0 in stage 1.0 (TID 7) in 111 ms on localbost (4/4)
16/04/16 16:30:53 INFO DAGScheduler: RegultStage 1 (collect at <console>:18) finished in 0.100 s
16/04/16 16:30:53 INFO DAGScheduler: Job 0 finished: collect at <console>:18, took 0.520306 s
res8: Array((String, Iterable[Int])] = Array((even,CompactBuffer(2, 4, 6, 8, 10, 12, 14)), (odd,CompactBuffer(1, 3, 5, 7, 9, 11, 13, 15)))
```

```
Example 15: groupByKey
```

```
\label{eq:val_name} $$ val name = sc.parallelize(List("kim","kumar","muthu","tim","lak","vams")) $$ val namekey = name.keyBy(\_.length) $$ val mycounter = namekey.map(x => (x._1,1)) $$ mycounter.collect
```

```
scala> val name = sc.parallelize[List("kim", "kumar", "nuthu", "tim", "lak", "vams"))
name: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[3] at parallelize at <console>:15

scala> val namekey = name.key8y( .length)
namekey: org.apache.spark.rdd.RDD[(Int, String)] = MapPartitionsRDD[4] at keyBy at <console>:17

scala> val mycounter = namekey.map{x => (x. 1,1))
nycounter: org.apache.spark.rdd.RDD[(Int, Int)] = MapPartitionsRDD[5] at map at <console>:19

scala> mycounter.collect
16/04/16 16:32:30 INFO SparkContext: Starting job: collect at <console>:22
16/04/16 16:32:30 INFO DAGScheduler: Got job 1 (collect at <console>:22) with 4 output partitions (allowLocal=false)
16/04/16 16:32:30 INFO DAGScheduler: Final stage: ResultStage 2(collect at <console>:22)
16/04/16 16:32:30 INFO DAGScheduler: Parents of final stage: List()
```

```
16/04/16 16:32:36 INFO Executor: Finished task 0.0 in stage 2.0 (TID 8). 750 bytes result sent to driver 16/04/16 16:32:30 INFO TaskSetManager: Finished task 0.0 in stage 2.0 (TID 8) in 24 ms on localhost (2/4) 16/04/16 16:32:30 INFO Executor: Finished task 3.0 in stage 2.0 (TID 11). 766 bytes result sent to driver 16/04/16 16:32:30 INFO TaskSetManager: Finished task 3.0 in stage 2.0 (TID 11) in 18 ms on localhost (3/4) 16/04/16 16:32:30 INFO Executor: Finished task 1.0 in stage 2.0 (TID 9). 766 bytes result sent to driver 16/04/16 16:32:30 INFO TaskSetManager: Finished task 1.0 in stage 2.0 (TID 9) in 22 ms on localhost (4/4) 16/04/16 16:32:30 INFO TaskSchedulerImpl: Removed TaskSet 2.0, whose tasks have all completed, from pool 16/04/16 16:32:30 INFO DAGScheduler: ResultStage 2 (collect at <console>:22) finished in 0.022 s 16/04/16 16:32:30 INFO DAGScheduler: Job 1 finished: collect at <console>:22, took 0.040170 s resl: Array((Int, Int)) = Array((3,1), (5,1), (5,1), (3,1), (4,1))
```

```
Example 16: Join

val name = sc.parallelize(List("kim","kumar","muthu","tim","lak","vams"))

val namekey = name.keyBy(_.length)

namekey.collect

val sub = sc.parallelize(List("English","Maths","Tamil","Science"))

val subkey = sub.keyBy(_.length)

subkey.collect

namekey.join(subkey).collect
```

```
scala> val name = sc.parallelize(List["kim", "Kunar", "nutho", "tim", "lak", "vams"))
name: org.apache.spark.rdd.RDD[string] = ParallelCollectionRDD[6] at parallelize at <console>:15

scala> val namekey = name.key8y( .length)
namekey: org.apache.spark.rdd.RDD[[Int, String]] = MapPartitionsRDD[7] at key8y at <console>:17

scala> namekey.collect
16/04/16 16:33:52 INFO SparkContext: Starting job: collect at <console>:20
16/04/16 16:33:52 INFO DAGScheduler: Got job 2 (collect at <console>:20) with 4 output partitions (allowLocal=false)
16/04/16 16:33:52 INFO DAGScheduler: Final stage: ResultStage 3(collect at <console>:20)

16/04/16 16:33:52 INFO Executor: Finished task 2.0 in stage 3.0 (TID 14). 764 bytes result sent to driver
16/04/16 16:33:52 INFO Executor: Finished task 3.0 in stage 3.0 (TID 15). 787 bytes result sent to driver
16/04/16 16:33:52 INFO TaskSetManager: Finished task 2.0 in stage 3.0 (TID 14) in 21 ms on localhost (3/4)
16/04/16 16:33:52 INFO TaskSetManager: Finished task 3.0 in stage 3.0 (TID 15) in 22 ms on localhost (4/4)
16/04/16 16:33:52 INFO TaskSetManager: ResultStage 3 (collect at <console>:20) finished in 0.032 s
16/04/16 16:33:52 INFO DAGScheduler: ResultStage 3 (collect at <console>:20) finished in 0.032 s
16/04/16 16:33:52 INFO DAGScheduler: Job 2 finished: collect at <console>:20, took 0.104835 s
res2: Array[{Int, String}] = Array((3,kim), (5,kumar), (5,muthu), (3,tim), (3,lak), (4,vams))
```

```
scala> val sub = sc.parallelize(List ("English", "Maths", "Tamil", "Science"))
sub: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[8] at parallelize at <console>:15

scala> val subkey = sub.key8y(_.length)
subkey: org.apache.spark.rdd.RDD[(Int, String)] = MapPartitionsRDD[9] at key8y at <console>:17

scala> subkey.collect
16/04/16 16:46:20 INFO SparkContext: Starting job: collect at <console>:20
16/04/16 16:46:20 INFO DAGScheduler: Got job 3 (collect at <console>:20) with 4 output partitions (allowLocal=false)
16/04/16 16:46:20 INFO DAGScheduler: Final stage: ResultStage 4(collect at <console>:20)
16/04/16 16:46:20 INFO DAGScheduler: Parents of final stage: List()

16/04/16 16:46:20 INFO TaskSetManager: Finished task 2.0 in stage 4.0 (TID 18) in 17 ms on localhost (2/4)
16/04/16 16:46:20 INFO TaskSetManager: Finished task 3.0 in stage 4.0 (TID 19) in 17 ms on localhost (3/4)
16/04/16 16:46:20 INFO TaskSetManager: Finished task 1.0 in stage 4.0 (TID 17). 766 bytes result sent to driver
16/04/16 16:46:20 INFO DAGScheduler: ResultStage 4 (collect at <console>:20) finished in 0.019 s
16/04/16 16:46:20 INFO DAGScheduler: ResultStage 4 (collect at <console>:20) finished in 0.019 s
16/04/16 16:46:20 INFO TaskSchedulerImpl: Removed TaskSet 4.0, whose tasks have all completed, from pool
```

16/04/16 16:46:20 INFO DAGScheduler: Job 3 finished: collect at <console>:20, took 0.035289 s
res3: Array[(Int, String)] = Array((7,English), (5,Maths), (5,Tamil), (7,Science))

```
scala> namekey.join(subkey).collect
16/04/16 16:47:29 INFO 5parkContext: Starting job: collect at <console>:24
16/04/16 16:47:29 INFO DAGScheduler: Registering RDO 7 (keyBy at <console>:17)
16/04/16 16:47:29 INFO DAGScheduler: Registering RDO 7 (keyBy at <console>:17)
16/04/16 16:47:29 INFO DAGScheduler: Got job 4 (collect at <console>:24) with 4 output partitions (allowLocal=false)
16/04/16 16:47:29 INFO DAGScheduler: Final stage: ResultStage 7(collect at <console>:24)
16/04/16 16:47:29 INFO DAGScheduler: Parents of final stage: List(ShuffleMapStage 5, ShuffleMapStage 6)

16/04/16 16:47:29 INFO TaskSetManager: Finished task 0.0 in stage 7.0 (TID 28) in 37 ms on localhost (2/4)
16/04/16 16:47:29 INFO Executor: Finished task 3.0 in stage 7.0 (TID 28) in 37 ms on localhost (2/4)
16/04/16 16:47:29 INFO Executor: Finished task 3.0 in stage 7.0 (TID 31). 882 bytes result sent to driver
16/04/16 16:47:29 INFO TaskSetManager: Finished task 3.0 in stage 7.0 (TID 31) in 57 ms on localhost (3/4)
16/04/16 16:47:29 INFO TaskSetManager: Finished task 3.0 in stage 7.0 (TID 29) in 60 ms on localhost (4/4)
16/04/16 16:47:29 INFO TaskSetManager: Finished task 1.0 in stage 7.0 (TID 29) in 60 ms on localhost (4/4)
16/04/16 16:47:29 INFO TaskSetManager: Removed TaskSet 7.0, whose tasks have all completed, from pool
16/04/16 16:47:29 INFO DAGScheduler: ResultStage 7 (collect at <console>:24) finished in 0.062 s
16/04/16 16:47:29 INFO DAGScheduler: Job 4 finished: collect at <console>:24, took 0.175332 s
res4: Array((Int, (String, String))] = Array((5, (kumar, Maths)), (5, (kumar, Tamil))), (5, (muthu, Maths)), (5, (muthu, Maths)), (5, (muthu, Tamil)))
```

```
Example 17: Join

Inner Join

val names1 = sc.parallelize(List("apple", "mango", "grapes")).map(a => (a,1))

val names2 = sc.parallelize(List("grapes", "litchi", "pears")).map(a => (a,1))

names1.join(names2).collect

leftOuterJoin

names1.leftOuterJoin(names2).collect

rightOuterJoin

names1.rightOuterJoin(names2).collect
```

```
scala> val names1 = sc.parallelize(List("apple", "mango", "grapes")).map(a => {a,1})
names1: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[14] at map at <console>:17

scala> val names2 = sc.parallelize(List("grapes", "litchi", "pears")).map(a => {a,1})
names2: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[16] at map at <console>:17

scala> names1.join(names2).collect
16/04/16 16:51:11 INFO parkContext: Starting job: collect at <console>:22
16/04/16 16:51:11 INFO DAGScheduler: Registering RDD 14 (map at <console>:17)
16/04/16 16:51:11 INFO DAGScheduler: Registering RDD 16 (map at <console>:17)
16/04/16 16:51:11 INFO DAGScheduler: Got job 5 (collect at <console>:22) with 4 output partitions (allowLocal=false)
16/04/16 16:51:11 INFO DAGScheduler: Final stage: ResultStage 10(collect at <console>:22)

16/04/16 16:51:11 INFO Executor: Finished task 3.0 in stage 10.0 (TID 43). 882 bytes result sent to driver
16/04/16 16:51:11 INFO Executor: Finished task 1.0 in stage 10.0 (TID 41). 882 bytes result sent to driver
16/04/16 16:51:11 INFO TaskSetManager: Finished task 1.0 in stage 10.0 (TID 41) in 31 ms on localhost (2/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 3.0 in stage 10.0 (TID 41) in 31 ms on localhost (3/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (3/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO TaskSetManager: Finished task 2.0 in stage 10.0 (TID 42) in 31 ms on localhost (4/4)
16/04/16 16:51:11 INFO Tas
```

```
scala> mames1.leftOuterJoin(names2).collect
16/84/16 16:52:16 INFO SparkContext: Starting job: collect at <console>:22
16/84/16 16:52:16 INFO DAGScheduler: Registering RDD 14 (map at <console>:17)
16/84/16 16:52:16 INFO DAGScheduler: Registering RDD 16 (map at <console>:17)
16/84/16 16:52:16 INFO DAGScheduler: Got job 6 (collect at <console>:22) with 4 output partitions (allowLocal=false)
16/84/16 16:52:16 INFO DAGScheduler: Final stage: ResultStage 13(collect at <console>:22) info DAGScheduler: Parents of final stage: List(ShuffleMapStage 12, ShuffleMapStage 11)

16/84/16 16:52:16 INFO DAGScheduler: Parents of final stage: List(ShuffleMapStage 12, ShuffleMapStage 11)

16/84/16 16:52:16 INFO ShuffleBlockFetcherIterator: Getting 3 non-empty blocks out of 4 blocks
16/84/16 16:52:16 INFO ShuffleBlockFetcherIterator: Started 8 remote fetches in 1 ms
16/84/16 16:52:16 INFO TaskSetManager: Finished task 1.0 in stage 13.0 (TID 53) in 38 ms on localhost (3/4)
16/84/16 16:52:16 INFO TaskSetManager: Finished task 2.0 in stage 13.0 (TID 54). 1131 bytes result sent to driver
16/84/16 16:52:16 INFO TaskSetManager: Finished task 2.0 in stage 13.0 (TID 54) in 40 ms on localhost (4/4)
16/84/16 16:52:16 INFO TaskSetManager: ResultStage 13 (collect at <console>:22) finished in 0.043 s
16/84/16 16:52:16 INFO DAGScheduler: ResultStage 13 (collect at <console>:22, took 0.323516 s
res6: Array[(String, (Int, Option[Int])]) = Array((grapes, (1, Some(1))), (apple, (1, None)), (mango, (1, None)))
```

```
Scala> names1.rightOuterJoin(names2).collect
16/04/16 10:53:35 INFO SparkContext: Starting job: collect at <console>:22
16/04/16 10:53:35 INFO MAGScheduler: Registering RDD 16 (map at <console>:17)
16/04/16 10:53:35 INFO DAGScheduler: Registering RDD 16 (map at <console>:17)
16/04/16 10:53:35 INFO DAGScheduler: Got job 7 (collect at <console>:22) with 4 output partitions (allowLocal=false)
16/04/16 10:53:35 INFO DAGScheduler: Final stage: ResultStage 16(collect at <console>:22)
16/04/16 10:53:35 INFO DAGScheduler: Parents of final stage: List(ShuffleMapStage 15, ShuffleMapStage 14)
16/04/16 10:53:35 INFO DAGScheduler: Missing parents: List(ShuffleMapStage 15, ShuffleMapStage 14)

16/04/16 10:53:35 INFO DAGScheduler: Missing parents: List(ShuffleMapStage 15, ShuffleMapStage 14)

16/04/16 10:53:35 INFO TaskSetManager: Finished task 3.0 in stage 10.0 (TID 67) in 23 ms on localhost (2/4)
16/04/16 10:53:35 INFO TaskSetManager: Finished task 1.0 in stage 10.0 (TID 68) in 26 ms on localhost (3/4)
16/04/16 10:53:35 INFO TaskSetManager: Finished task 0.0 in stage 10.0 (TID 64). Ill5 bytes result sent to driver
16/04/16 10:53:35 INFO TaskSetManager: Finished task 0.0 in stage 10.0 (TID 64). Ill5 bytes result sent to driver
16/04/16 10:53:35 INFO TaskSetManager: ResultStage 16.0 (TID 64). Ill5 bytes result sent to driver
16/04/16 10:53:35 INFO DAGScheduler: ResultStage 16 (collect at <console>:22) finished in 0.032 s
16/04/16 16:53:35 INFO DAGScheduler: Job 7 finished: collect at <console>:22, took 0.109093 s
16/04/16 16:53:35 INFO DAGScheduler: Job 7 finished: collect at <console>:22, took 0.109093 s
16/04/16 10:53:35 INFO DAGScheduler: Job 7 finished: collect at <console>:22, took 0.109093 s
```



# Exercise 4: Operations On Multiple RDDs

#### Overview

In this lab, we will look at several transformations and examine the optimizations and visualise with DAG.

#### Builds on

Previous labs for the transformations we'll use.

#### Run time

30-40 minutes

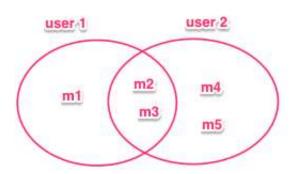
## Meetups: Our data for this lab

For this lab, assume that we have users who are attending several meetups. We will start with two users, who are attending the following meetups:

User1 attends meetups: m1, m2 and m3. User2 attends meetups: m2, m3, m4 and m5.

Each user's meetups will be in separate RDDs, so you'll have two RDDs to work with. We'll analyze the data for the two users by performing operations over both RDDs. We'll also look at the Spark Shell UI to get an idea of how Spark is processing the data.

Operations that you work with will include union, intersection, distinct, and subtract.



#### \*\*\* Create the RDDs \*\*\*

scala> val u1 = sc.parallelize(List("m1", "m2", "m3"))

u1: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[0] at parallelize at <console>:24

scala> val u2 = sc.parallelize(List("m2", "m3", "m4", "m5"))

u2: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[1] at parallelize at <console>:24

### -- Once you've done that, look at the Jobs tab of the UI - do you see anything?

You won't see any jobs yet. There have been no actions, so Spark is being lazy, and not doing anything yet.

#### \*\*\* Using your two RDDs, find meetups common to both users

scala> val common = u1.intersection(u2)

common: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[7] at intersection at <console>:28

scala> common.collect

res0: Array[String] = Array(m2, m3)

#### \*\*\* Find meetups attended by either user1 or user2.

scala> val either = u1.union(u2).distinct()

either: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[11] at distinct at <console>:28

scala> either.collect

res1: Array[String] = Array(m1, m2, m3, m4, m5)

#### \*\*\* Find meetups for each user that only one attended

#### -- Find meetups that ONLY u1 attended (That is, u1 attended, but u2 did not.)

scala> val onlyU1 = u1.subtract(u1.intersection(u2))

onlyU1: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[21] at subtract at <console>:28

scala> onlyU1.collect

res2: Array[String] = Array(m1)

-- Find meetups that ONLY u2 attended (That is, u2 attended, but u1 did not.)

scala> val onlyU2 = u2.subtract(u1.intersection(u2))

onlyU2: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[31] at subtract at <console>:28

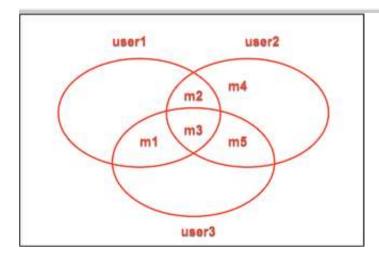
scala> onlyU2.collect

res3: Array[String] = Array(m5, m4)

- \*\*\* Find recommendations based on the requirements in the lab.
- \* A user should not be attending a meetup to be recommended.
- \* A meetup should be in attended by both other users to be recommended.

Let's introduce user3 [1]

User3 attends meetups: m1, m3, and m5, as illustrated below.



Based on this, we would recommend the following to the users:

u1: m5 u2: m1

u3: m2

## **Tasks**

- reate an RDD (u3) with the meetups it is attending (m1, m3, m5).
- ➤ Using your three RDDs, find the recommendations for u1 based on the rules above.
  - Look at the diagram above, and put into words how you would compute the recommendations.
  - Once you have a clear idea of how to do it, perform the RDD operations to

accomplish it in the Spark Shell.

• You'll need to use all 3 RDDs in your transformation for this. Look at the job DAG in the UI after you've done this.

#### -- Create u3

```
scala> val u3 = sc.parallelize(List("m1", "m3", "m5"))
```

u3: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[32] at parallelize at <console>:24

#### -- Recommendations for u1:

```
scala> val forU1 = u2.intersection(u3).subtract(u1)
```

forU1: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[42] at subtract at <console>:30

```
scala> forU1.collect
```

res4: Array[String] = Array(m5)

#### -- Recommendations for u2:

```
scala> val forU2 = u1.intersection(u3).subtract(u2)
```

forU2: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[52] at subtract at <console>:30

scala> forU2.collect

res5: Array[String] = Array(m1)

#### -- Recommendations for u3:

```
scala> val forU3 = u1.intersection(u2).subtract(u3)
```

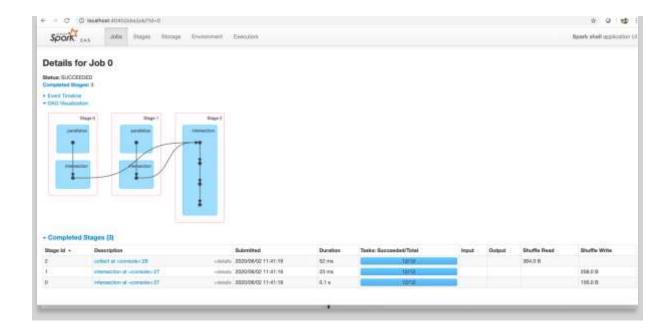
forU3: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[62] at subtract at <console>:30

scala> forU3.collect

res6: Array[String] = Array(m2)

#### Discussions on What's Seen in the Spark Shell UI

Let's consider at some of the results you might have seen earlier in the ULEE Here's the DAG from finding the meetups in common (an intersection of the two RDDs).

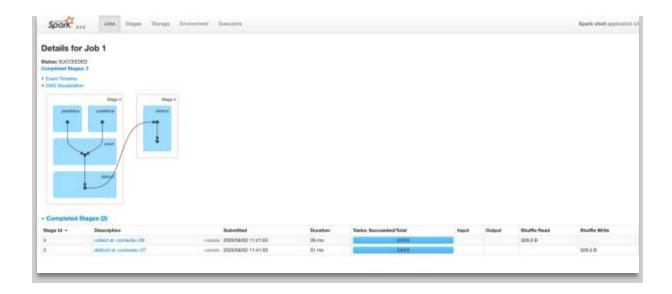


What you're seeing is this.

- Each shaded blue box represents a user operation.
- A dot in the box represents an RDD created in the corresponding operations.
- ➤ Operations are grouped by stages (In a stage, operations on partitions are pipelined in the same task).
- You can click on any stage, to drill down into it for more detail.
- ➤ Parallelization of each RDD occurs in one stage (e.g. on one node, with local data)
- > Some of the intersection can happen on one stage (using whatever data is local)
- ➤ Some of the intersection happens in another stage
  - Because it can no longer be done with local data it involves data distributed over the cluster.
  - Data is **shuffled** for the intersection to be done (i.e. sent from one node to another).
  - Shuffling is expensive we'll talk more about this later.

Here are the details on the stages (from the **Completed Stages** section on the same page). It gives details on the data that is shuffled.

Here's another DAG - from finding the meetups attended by **either** user (a union of the two RDDs).



What you're seeing is this.

- > Parallelization of each RDD occurs in one stage (e.g. on one node, with local data)
  - And yes, Spark has to parallelize again, because default is not to cache an RDD.
- > The union happens in the same stage It can all be done with local data
  - Part of the distinct has to be done in a separate stage it also requires shuffling data



# Exercise5: Building Spark Application using IntelliJ IDE

#### Overview

In this lab, we will look at several transformations and examine the optimizations that Catalyst performs.

#### Builds on

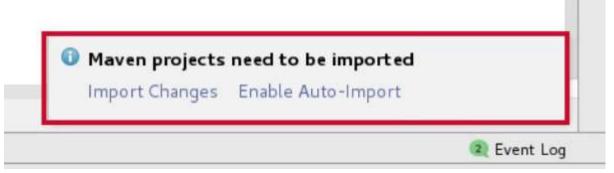
Previous labs for the transformations we'll use.

#### Run time

20-30 minutes

Goal: Create a scala project for Spark. Use maven for lib management.

- 1. Launch IntelliJ using the link on your lab environment desktop.
- 2. Click Open to open a project.
- 3. Locate and select the project's root folder (Lesson4, which you downloaded earlier) and click
- 4. OK. Wait while the project loads and the workspace is prepared.
- 5. If there's a "Maven projects need to be imported" popup message in the lower-right corner, click



- 6. When you have finished the code, build the JAR file which will be submitted to Spark. To do this with IntelliJ:
  - a. Navigate to View > Tool Windows > Maven Projects. This opens a Maven Projects
  - pane on the right side of the screen. You will see sfpdapp listed as a project:
- 7. Now add a new Scala object file "AirportsByLatitude" to project.

**Use Case**: Create a Spark program to read the airport data from airports.text, find all the airports whose latitude are bigger than 40.

Then output the airport's name and the airport's latitude to airports by latitude.

Each row of the input file contains the following columns:

# Airport ID, Name of airport, Main city served by airport, Country where airport is located, IATA/FAA code,

ICAO Code, Latitude, Longitude, Altitude, Timezone, DST, Timezone in Olson format

#### Sample output:

"St Anthony", 51.391944 "Tofino", 49.082222

```
import org.apache.spark.{SparkConf, SparkContext}

object AirportsByLatitude{
    def main(args: Array[String]) {
      val conf = new SparkConf().setAppName("airports").setMaster("local[2]")
      val sc = new SparkContext(conf)
      val airports = sc.textFile("<hdfs path>/airports.text")
      val airportsInUSA = airports.filter(line => line.split(Utils.COMMA_DELIMITER)(6).toFloat > 40)
      val airportsNameAndCityNames = airportsInUSA.map(line => {
       val splits = line.split(Utils.COMMA_DELIMITER)
            splits(1) + ", " + splits(6)
      })
      airportsNameAndCityNames.saveAsTextFile("<hdfs path>/airports_by_latitude.text")
    }
}
```

The code is already imported into IntelliJ in TrainingSpark workspace.

Execute the code, Right click -> Run as Scala Application. It has created the output folder with the desired output. You can go and verify the output.



#### Use Case:

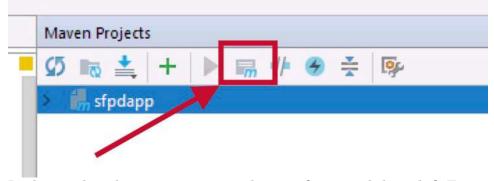
Create a Spark program to read the airport data from airports.text, find all the airports which are located in United States and output the airport's name and the city's name to out/airports\_in\_usa.

Each row of the input file contains the following columns:

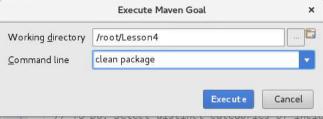
Airport ID, Name of airport, Main city served by airport, Country where airport is located, IATA/FAA code, ICAO Code, Latitude, Longitude, Altitude, Timezone, DST, Timezone in Olson format

#### Sample output:

- "Putnam County Airport", "Greencastle"
- "Dowagiac Municipal Airport", "Dowagiac"
- 8. Click the Execute Maven Goal icon in the top bar of the Maven Projects window:



In the window that appears, enter clean package, and then click Execute:



- 9. When it has finished downloading repositories and packaging the project, you will see a BUILD SUCCESS message in the status window. This may take a few minutes.
- 10. Open a terminal window on the host, and

11. Take this jar and submit using spark-submit command

```
cd /<path>/Lesson4/target

Navigate to the folder where Spark is installed.

$ bin/spark-submit --master yarn --class solution.AirportsByLatitude
/<path>/name of the jar
```

Next verify the output in the output directory of hdfs.

```
(base) [cloudera@quickstart spark-2.4.4-bin-hadoop2.6]$ hdfs dfs -ls /user/cloudera/airports_by_latitude
Found 2 items
-rw-r--r-- 3 cloudera cloudera 0 2019-11-16 02:06 /user/cloudera/airports_by_latitude/_SUCCESS
-rw-r--r-- 3 cloudera cloudera 92308 2019-11-16 02:06 /user/cloudera/airports_by_latitude/part-00000
(base) [cloudera@quickstart spark-2.4.4-bin-hadoop2.6]$ ■
```



# Exercise 6: Accumulators

Here we have the results of stack overflow annual salary survey for developers worldwide. This is a subset of the survey results that's open to CSV via an Excel. The first row is the header we are mostly interested in the third column which is the country of the developer and the 15th column which is the salary made a point.

But as you see we have some responses which don't have the salary midpoint records here. We want to answer several questions.

- First how many records do we have in the survey result.
- > Second how many records are missing the salary middle point third.
- ➤ How many records are from Canada.

But we want to answer all those questions by passing over all the data only once.

Let us create a Scala class as StackOverFlowSurvey in the package com.sparkTutorial.advanced.accumulator with the below code.

```
package com.sparkTutorial.advanced.accumulator
import com.sparkTutorial.commons.Utils
import org.apache.log4j.{Level, Logger}
```

```
import org.apache.spark.{SparkConf, SparkContext}
object StackOverFlowSurvey {
  def main(args: Array[String]) {
    Logger.getLogger("org").setLevel(Level.ERROR)
    val conf = new
SparkConf().setAppName("StackOverFlowSurvey").setMaster("local[1]")
    val sparkContext = new SparkContext(conf)
    val total = sparkContext.longAccumulator
    val missingSalaryMidPoint = sparkContext.longAccumulator
    val responseRDD = sparkContext.textFile("in/2016-stack-overflow-
survey-responses.csv")
    val responseFromCanada = responseRDD.filter(response => {
      val splits = response.split(Utils.COMMA DELIMITER, -1)
      total.add(1)
      if (splits(14).isEmpty) {
        missingSalaryMidPoint.add(1)
      splits(2) == "Canada"
    })
    println("Count of responses from Canada: " +
responseFromCanada.count())
    println("Total count of responses: " + total.value)
    println("Count of responses missing salary middle point: " +
missingSalaryMidPoint.value)
```



# Exercise 7: Broadcast variables

Here we have the maker space data across the UK. The first row is the header. This data set includes the name of the maker space email address postcard number of visitors etc. We want to answer the following question.

➤ How are those makers space distributed across different regions in the UK.

However, we only have the postcode for each maker space.

We don't know the region of each maker's space to answer the above question.

We need to introduce another dataset here.

We also have the UK postcode data for each postcode prefix we can find out which region it belongs to.

Combining those two data sets we should be able to answer the question how are those makers spaces distributed across different regions in the UK.

Let us create a Scala class as UkMakerSpaces in the package "com.sparkTutorial.advanced.broadcast" and paste the below code:

```
package com.sparkTutorial.advanced.broadcast
import com.sparkTutorial.commons.Utils
import org.apache.log4j.{Level, Logger}
import org.apache.spark.{SparkConf, SparkContext}
import scala.io.Source
object UkMakerSpaces {
  def main(args: Array[String]) {
    Logger.getLogger("org").setLevel(Level.ERROR)
    val conf = new
SparkConf().setAppName("UkMakerSpaces").setMaster("local[1]")
    val sparkContext = new SparkContext(conf)
    val postCodeMap = sparkContext.broadcast(loadPostCodeMap())
    val makerSpaceRdd = sparkContext.textFile("in/uk-makerspaces-
identifiable-data.csv")
    val regions = makerSpaceRdd
      .filter(line => line.split(Utils.COMMA DELIMITER, -1)(0) !=
"Timestamp")
      .filter(line => getPostPrefix(line).isDefined)
      .map(line =>
postCodeMap.value.getOrElse(getPostPrefix(line).get, "Unknown"))
    for ((region, count) <- regions.countByValue()) println(region +</pre>
" : " + count)
```

```
def getPostPrefix(line: String): Option[String] = {
   val splits = line.split(Utils.COMMA_DELIMITER, -1)
   val postcode = splits(4)
   if (postcode.isEmpty) None else Some(postcode.split(" ")(0))
}

def loadPostCodeMap(): Map[String, String] = {
   Source.fromFile("in/uk-postcode.csv").getLines.map(line => {
     val splits = line.split(Utils.COMMA_DELIMITER, -1)
        splits(0) -> splits(7)
   }).toMap
}
```



# Exercise 8: Dataframe, Datasets and Spark SQL

Two ways to create Spark Dataframes:

Creating DataFrames from Existing RDD:

- 1. Infer schema by Reflection
  - a. Convert RDD containing case classes.
  - b. Use when schema is known.
- 2. Construct schema programmatically
  - a. Schema is dynamic
  - b. Number of fields in case class is more than 22 fields.

#### TWO WAYS TO DEFINE A SCHEMA

#### 1. Define Schema Programmatically

```
import org.apache.spark.sql.types._
import org.apache.spark.sql.functions.{col, expr}
val jsonFile = "file:///home/cloudera/Desktop/SW/in/Data/blogs.json"
 // Define our schema programmatically
 val schema = StructType(Array(StructField("Id", IntegerType, false),
  StructField("First", StringType, false),
  StructField("Last", StringType, false),
   StructField("Url", StringType, false),
   StructField("Published", StringType, false),
  StructField("Hits", IntegerType, false),
  StructField("Campaigns", ArrayType(StringType), false)))
/Create a DataFrame by reading from the JSON file a predefined Schema
  val blogsDF = spark.read.schema(schema).json(jsonFile)
  //show the DataFrame schema as output
  blogsDF.show(truncate = false)
  // print the schemas
```

```
print(blogsDF.printSchema)
  print(blogsDF.schema)
  // Show columns and expressions
  blogsDF.select(expr("Hits") * 2).show(2)
  blogsDF.select(col("Hits") * 2).show(2)
  blogsDF.select(expr("Hits * 2")).show(2)
 // show heavy hitters
 blogsDF.withColumn("Big Hitters", (expr("Hits > 10000"))).show()
// Concatenate three columns, create a new column, and show the
// newly created concatenated column
blogsDF.withColumn("AuthorsId".
                                         (concat(expr("First"),
                                                                    expr("Last"),
expr("Id")))).select(col("AuthorsId")).show(4)
// Sort by column "Id" in descending order
blogsDF.sort(col("Id").desc).show()
blogsDF.sort($"Id".desc).show()
// In Scala to save as a Parquet file
val parquetPath = <path>
blogsDF.write.format("parquet").save(parquetPath)
```

Note: the expressions blogs\_df.sort(col("Id").desc) and blogsDF.sort(\$"Id".desc) are identical. They both sort the DataFrame column named Id in descending order: one uses an explicit function, col("Id"), to return a Column object, while the other uses \$ before the name of the column, which is a function in Spark that converts column named Id to a Column.

**Assignment 1**: let's read a large CSV file containing data on San Francisco Fire Department calls. we will define a schema for this file and use the DataFrameReader class and its methods to tell Spark what to do. Because this file contains 28 columns and over 4,380,660 records,2 it's more efficient to define a schema than have Spark infer it.

#### Hint:

val sfFireFile="/path/sf-fire-calls.csv"

```
val fireDF = spark.read.schema(fireSchema)
.option("header", "true")
.csv(sfFireFile)
```

- What were all the different types of fire calls in 2018?
- What months within the year 2018 saw the highest number of fire calls?
- Which neighborhood in San Francisco generated the most fire calls in 2018?
- Which neighborhoods had the worst response times to fire calls in 2018?
- Which week in the year in 2018 had the most fire calls?
- Is there a correlation between neighborhood, zip code, and number of fire calls?
- How can we use Parquet files or SQL tables to store this data and read it back?

#### 2. Infer schema by Reflection

let's look at a collection of readings from Internet of Things (IoT) devices in a JSON file (we use this file in the end-to-end example later in this section).

Let us create Case class for the file

```
case class DeviceIoTData (battery_level: Long, c02_level: Long, cca2: String, cca3: String, cn: String, device_id: Long, device_name: String, humidity: Long, ip: String, latitude: Double, lcd: String, longitude: Double, scale:String, temp: Long, timestamp: Long)

val ds = spark.read.json("file:///home/cloudera/Desktop/SW/in/Data/iot_devices.json").as[DeviceIoTData]

val filterTempDS = ds.filter({d => {d.temp > 30 && d.humidity > 70})}

filterTempDS.show(5, false)
```

```
case class DeviceTempByCountry(temp: Long, device_name: String, device_id:
Long,
 cca3: String)
val dsTemp = ds
 .filter(d \Rightarrow {d.temp > 25})
 .map(d => (d.temp, d.device_name, d.device_id, d.cca3))
 .toDF("temp", "device_name", "device_id", "cca3")
 .as[DeviceTempByCountry]
dsTemp.show(5, false)
Alternatively, you could express the same query using column names and then
cast to a Dataset[DeviceTempByCountry]:
// In Scala
val dsTemp2 = ds
 .select($"temp", $"device_name", $"device_id", $"device_id", $"cca3")
 .where("temp > 25")
 .as[DeviceTempByCountry]
```

**Assignment 2**: Using the Dataset API, try to execute to do the following:

- Detect failing devices with battery levels below a threshold.
- Identify offending countries with high levels of CO2 emissions.
- Compute the min and max values for temperature, battery level, CO2, and humidity.
- Sort and group by average temperature, CO2, humidity, and country.

#### Using Spark SQL with queries

We will here work on the Airline On-Time Performance and Causes of Flight Delays data set, which contains data on US flights including date, delay, distance, origin, and destination. It's available as a CSV file with over a million records. Using a schema, we'll read the data into a DataFrame and register the DataFrame as a temporary view (more on temporary views shortly) so we can query it with SQL.

```
// Path to data set
val csvFile="/home/cloudera/Desktop/SW/in/Data/departuredelays.csv"

// Read and create a temporary view
// Infer schema (note that for larger files you may want to specify the schema)
val df = spark.read.format("csv").option("inferSchema", "true").option("header", "true").load(csvFile)
// Create a temporary view
df.createOrReplaceTempView("us_delay_flights_tbl")
```

Now that we have a temporary view, we can issue SQL queries using Spark SQL. These queries are no different from those you might issue against a SQL table in, say, a MySQL or PostgreSQL database. The point here is to show that Spark SQL offers an ANSI:2003–compliant SQL interface, and to demonstrate the interoperability between SQL and DataFrames.

The US flight delays data set has five columns:

- The date column contains a string like 02190925. When converted, this maps to 02-19 09:25 am.
- The delay column gives the delay in minutes between the scheduled and actual departure times. Early departures show negative numbers.
- The distance column gives the distance in miles from the origin airport to the destination airport.
- The origin column contains the origin IATA airport code.
- The destination column contains the destination IATA airport code.

With that in mind, let's try some example queries against this data set.

First, we'll find all flights whose distance is greater than 1,000 miles:

spark.sql("""SELECT distance, origin, destination FROM us\_delay\_flights\_tbl WHERE distance > 1000 ORDER BY distance DESC""").show(10)

➤ Next, we'll find all flights between San Francisco (SFO) and Chicago (ORD) with at least a two-hour delay:

spark.sql("""SELECT date, delay, origin, destination FROM us\_delay\_flights\_tbl WHERE delay > 120 AND ORIGIN = 'SFO' AND DESTINATION = 'ORD' ORDER by delay DESC""").show(10)

- ➤ **Assignment**: It seems there were many significantly delayed flights between these two cities, on different dates. (As an exercise, convert the date column into a readable format and find the days or months when these delays were most common. Were the delays related to winter months or holidays?)
- Let's try a more complicated query where we use the CASE clause in SQL. In the following example, we want to label all US flights, regardless of origin and destination, with an indication of the delays they experienced: Very Long Delays (> 6 hours), Long Delays (2–6 hours), etc. We'll add these human-readable labels in a new column called Flight\_Delays:



#### Overview

In this lab, we will look at several transformations and optimizations on Dataframes and Datasets.

#### Builds on

Previous labs for the transformations we'll use.

#### Run time

20-30 minutes

# > Explore the Spark shell UI Tasks

Open the FireFox browser (In cloudera dekstop, **Applications Menu | FireFox**), and go to the launch **Spark\_UI (localhost:4040)** 

In the page that opens, look for the application [5]

That should open the Spark UI Below is a sample screen shot of the UI





**NOTE:** In all screen shots, you might see differences in job ids, stage ids and other details that depend on the exact state of your Spark shell. This is because the shell used for the screen shots is likely to have executed a different overall sequence of transformations than your shell since it was started. This is not important - the important details should be easily recognizable.

# Lab: 10.1

#### Spark SQL Basics

#### Overview

In this lab, we will use some of the basic functionality of Spark SQL, including:

Read and write data files in varying formats Create DataFrames

#### Run time

15-20 minutes

#### Lab Preparation

#### **Tasks**

Load data from the following files, if you haven't already

- people.json as JSON
- *wiki-pageviews.txt* as text
- github.json as JSON

```
//Scala Code
***

*** Load Text and JSON Data ***

***

Read people.json
> val folksDF=spark.read.json("data/people.json")

Read wiki-pageviews.txt
> var viewsDF = spark.read.text("data/wiki-pageviews.txt")

Display some of the data
> folksDF.limit(5).show
> viewsDF.limit(5).show

***

*** Read More Complex Data ***

***

Read/display data in github.json

> val githubDF=spark.read.json("data/github.json")
> githubDF.limit(5).show

***

***

Write Data ***

***

***

Write Data ***

***
```

```
Write out folksDF as parquet file
> folksDF.write.parquet("data/people.parquet")

Write out folksDF as a CSV file
> folksDF.coalesce(1).write.option("header", true).csv("data/people.csv")
```

#### View the Schema

```
scala> githubDF.printSchema()
   - actor: struct (nullable = true)
        -- avatar_url: string (nullable = true)
       -- gravatar_id: string (nullable = true)
       -- id: long (nullable = true)
       -- login: string (nullable = true)
       -- url: string (nullable = true)
    - created_at: string (nullable = true)
   -- id: string (nullable = true)
-- org: struct (nullable = true)
        -- avatar_url: string (nullable = true)
       -- gravatar_id: string (nullable = true)
       -- id: long (nullable = true)
       -- login: string (nullable = true)
       -- url: string (nullable = true)
   -- payload: struct (nullable = true)
        -- action: string (nullable = true)
        -- before: string (nullable = true)
        -- comment: struct (nullable = true)
              - _links: struct (nullable = true)
                  -- html: struct (nullable = true)
                      |-- href: string (nullable = true)
                  -- pull_request: struct (nullable = true)
                      |-- href: string (nullable = true)
                  -- self: struct (nullable = true)
                      |-- href: string (nullable = true)
              - body: string (nullable = true)
            |-- commit_id: string (nullable = true)
            |-- created_at: string (nullable = true)
            |-- diff_hunk: string (nullable = true)
            |-- html_url: string (nullable = true)
```

The githubDF schema is quite complex. We'll look at some ways of dealing with it soon.

# Declare a Schema Explicitly Tasks

Declare a schema for folksDF. We illustrate this below. We also supply this code in a below snippet - *peopleSchema.txt*:

The file is stored in the Datasets.

```
import org.apache.spark.sql.types._
```

val mySchema = (new StructType).add("name", StringType).add("gender",
StringType).add("age", IntegerType)

Read data/people.json again, but this time supply the schema you declared.

> var folksDF = spark.read.schema(mySchema).json("data/people.json")

# # Python

from pyspark.sql.types import \*

mySchema =

StructType().add("name",StringType()).add("gender",StringType()).add("age",S

folksDF = spark.read.schema(mySchema).json("data/people.json")

# Lab 10.2

#### **DataFrame Transformations**

#### Overview

In this lab, we will perform a number of transformations on dataframes. This will provide exposure to the API, and some experience in using it.

We will cover a number of common techniques, but will not attempt to cover all

of the API, or try to illustrate every possible technique. The API is too large for that.

We'll be working at the Spark shell for all of these.

#### Run time

30-40 minutes

#### Assignment:

- > Display the data. Filter on age and display the data.
- > Filter on gender and display the data.
- > Count how many "F" and "M" items there are.
- ➤ [Optional] Find the oldest person with gender "F".
  - Use SQL to write this query e.g. spark.sql(" ... ")
  - Remember to register a temporary table.
  - Use a subquery to find the maximum age.

```
Solution:

***

*** Simple Transformations ***

***
```

```
Read people.json if you haven't already
> val folksDF=spark.read.json("data/people.json")
Display the data.
> folksDF.show
Filter on age and display the data
Filter on gender and display the data
> folksDF.filter('age>25).show // With age greater than 25
+---+
|age|gender|name|
+---+
| 35| M|John|
| 40| F|Jane|
| 52 | F | Sue |
+---+
> folksDF.filter('age>25 && 'age<50).show // With age greater than 25 and less
than 50
+---+
|age|gender|name|
+---+
| 35| M|John|
| 40| F|Jane|
+---+
```

```
> folksDF.filter('gender === "F").show
+---+
|age|gender|name|
+---+
| 40| F|Jane|
| 52 | F | Sue |
+---+
Count how many "F" and "M" items there are.
> folksDF.groupBy('gender).count.show
+----+
|gender|count|
+----+
  F|2|
   M \mid 2 \mid
+----+
[Optional] Find the oldest person with gender "F"
> folksDF.createOrReplaceTempView("people")
> spark.sql("SELECT name, age FROM people WHERE gender = 'F' and age =
(select max(age) FROM people WHERE gender='F')").show
+---+
|name|age|
+---+
| Sue | 52 |
```

+---+

#### Working with More Complex Data

#### Tasks

Read *github.json* if you haven't already

#### > val githubDF=spark.read.json("data/github.json")

- Look at the schema and 5 sample rows (limit(5).show) again.
- Look at the githubDF schema and a few rows (limit().show) again.
- > Select the login value of the actor and display it.
- Find out how many unique logins there are in the data.
- ➤ Each row in this data, contains a type column. Display all the unique values for the 'type' column.
- ➤ Determine how may rows have a type of CreateEvent

#### > githubDF.limit(5).show

Select the actor column and view it and its schema

- > githubDF.limit(5).select('actor).show(false)
- > githubDF.select('actor).printSchema

Select the login value of the actor and display it

> actorDF.select("actor.login").limit(5).show

Find out how many unique logins there are in the data.

> actorDF.select("actor.login").dropDuplicates.count

Retrieve all the unique values for the 'type' column.

- > gitHubDF.dropDuplicates("type").count
- > gitHubDF.dropDuplicates("type").show

#### Working with Text Data

#### Tasks

- Read *wiki-pageviews.txt* if you haven't already
- ➤ Show 5 rows from this data so you can examine it.

val viewsDF=spark.read.text("<path>/wiki-pageviews.txt")

viewsDF.limit(5).show

- > We can see that each row contains a single string.
  - Unlike JSON data, there is no automatic application of a schema.
  - This is a cumbersome format to work with, and we'll see ways to transform it into an easier to use format later.

#### Summary

We've practiced using some of the common transformations that are available for dataframes. We'll continue using this data to delve into Spark's capabilities.



# Lab 10.3: The Dataset Typed API

#### Overview

In this lab, we will work with the Dataset typed API which provides compiletime type safety.

Continue working in your spark shell in this lab as previously. You'll work with some of the dataframes created in the previous lab.

#### What About Python?

Python does NOT have Dataset. It is a dynamically typed language, and thus the strongly typed Dataset makes no sense for it.

#### Run time

30-40 minutes

#### Music Data File

We've provided a small and simple JSON data file (*<path>/data/music.json*) containing data about music items. We'll use it to do our Dataset work in this lab. It has four pieces of data in it:

- title: Title of the item.
- artist: Artist who released the item.
- price: Price of the item[sep]
- category: The music category (e.g. Pop)

#### **Tasks**

- > Create a DataFrame by reading in the music data in *data/music.json*
- > View the schema of this DataFrame.
- ➤ Display the data in the DataFrame (show).

#### Simple Dataset Usage Tasks

- ➤ Define a MusicItem case class suitable for our *music.json* data. Using this class, create a typed Dataset from the musicDF dataframe created
- > with this data in the previous lab.
- ➤ What type is musicDS? Display the data in the dataset.



#### Compare: DataFrame vs. Dataset

We'll perform some operations on our DataFrame and Dataset representations, to get a feel of the difference.

#### Tasks

- > Filter on category
  - Using musicDF get all the items in the "Pop" category.
  - Using musicDS get all the items in the "Pop" category.
- > Get lowest priced item in each category.
  - You'll need to group, then do an aggregation.
  - Using musicDF and untyped transformations, get the lowest price item in a category.
  - Using musicDS and typed transformations, get the lowest price item in a category.
- ransform data so that the price is reduced 10%. (You can do this by multiplying it by 0.9).
  - Using musicDF (the DataFrame) transform it to a DataFrame where the price is reduced by 10%.
    - Hint use a select and a literal to create the new price values.
  - Using musicDS (the Dataset) transform it to a Dataset where the price is doubled.
    - o Use map()
    - o Make sure you know how your case class is defined (i.e. what order the
    - o fields appear).
- Make a mistake watch how errors are caught.
  - Using musicDF (the DataFrame) make a mistake in transforming it when you modify the price.
    - o Try multiplying the price by a literal string e.g. lit("A")
    - What happens?
  - Using musicDS (the Dataset) make a mistake in transforming when you modify the price.
    - o Try multiplying the price by a string (e.g. "A")
    - What happens?

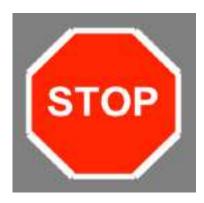
# Solution: \*\*\* \*\*\* Compare: DataFrame vs. DataSet \*\*\* \*\*\*

```
-- Filter on category.
> musicDF.filter('category === "Pop").show
> musicDS.filter(mi => mi.category == "Pop").show
- Using musicDF and untyped transformations, get the lowest price item
in a category.
> musicDF.groupBy('category).min("price").show
>musicDS.groupByKey(mi=>
mi.category).agg(min('price).as[Double]).show
- Transform data so that the price is reduced 10%. (You can do this by
multiplying it by 0.9).
// DataFrame
> musicDF.select('title, 'artist, 'category, 'price*lit(0.9)).show
-- Make a mistake - watch how errors are caught.
// DataFrame - error not caught - get erroneous results.
a> musicDF.select('title, 'artist, 'category, 'price*lit("A")).show
// DataSet - error caught at parsing time.
> musicDS.map (mi => MusicItem(mi.title, mi.artist, mi.category,
mi.price*"A")).show
> musicDS.map (mi => MusicItem(mi.title, mi.artist, mi.category,
mi.price*"A")).show
```

#### Summary

We've practiced some fairly simple transformations with Datasets. We've also done the same thing with DataFrames to get a feel of the di!erence.

DataFrames are a little easier to program to. However, as we saw, Datasets can catch errors earlier than DataFrames. This can be important when debugging and maintaining a large system.



# Lab 10.4: Splitting Text Data

#### Overview

In this lab, we'll work with text data. The data is regularly structured, but since it's in text format, Spark can't deduce the structure on its own.

We'll use some of the DataFrame tools to create a DataFrame with a schema that's easy to use for querying the data.

The data we'll use contains page view data from Wikimedia.

#### Run time

30-40 minutes

#### Wikimedia PageView Data File

We provide a data file (*<path>/data/wiki-pageviews.txt*) that contains a dump of pageview data for many projects under the Wikimedia umbrella.

The data file has lines containing four fields.

- 1. Domain.project (e.g. "en.b") [SEP]
- 2. Page name (e.g. "AP\_Biology/Evolution") 🔀
- 3. Page view count (e.g. 1)
- 4. Total response size in bytes (e.g. 10662 but for this particular dump, value is always 0). [5]
  - The data is in simple text format, so when we read it in, we get a dataframe with a single column a string containing all the data in each row. This is cumbersome to work with. In this lab, we'll apply a better schema to this data.
- > Tasks Create a DataFrame by reading in the page view data in spark-labs/data/wiki-pageviews.txt.

- Once you've created it, view a few lines to see the format of the data.
  - o You'll see that you have one line of input per dataframe row.

#### Split the Lines

Our first step in creating an easier to use schema is splitting each row into separate columns. We'll use the split() function defined in the Spark SQL funtions. We've used this in our word count examples in the main manual.

#### Tasks

- > Create a dataframe by splitting each line up. [1]
  - Use split() to split on a whitespace (pattern of "\\s+")

**Python:** You'll need to import from the functions module as shown in the Python example below

```
//Python
> from pyspark.sql.functions import *
> splitViewsDF = // ...
```

#### Create a Better Schema

We'll create a dataframe with an easier-to-use schema containing the following columns which align with the data in the views file.

- domain: String The domain.project data.
- pageName: String The page name.
- > viewCount: Integer The view count.
- > size: Long The response size (always 0 in this file, but we'll keep it in our dataframe).

Execute the below queries one after the other

```
***

*** Wikimedia PageView Data File ***

***

Read the data file, and show some lines

> val viewsDF=spark.read.text("data/wiki-pageviews.txt")

> viewsDF.limit(5).show
```

```
***
*** Split the Lines ***
Split on whitespace, show some lines, show the schema
> val splitViewsDF = viewsDF.select(split('value, "\\s+").as("splitLine"))
> splitViewsDF.limit(5).show(false)
> splitViewsDF.printSchema
***
*** Create a Better Schema ***
***
Create a dataframe with a better schema, view the schema, and view some
data.
>val
                         viewsWithSchemaDF
splitViewsDF.select('splitLine(0).as("domain"),
'splitLine(3).cast("long").as("size"))
> viewsWithSchemaDF.printSchema
> viewsWithSchemaDF.limit(5).show
*** Try Some Queries ***
Find rows where the viewCount > 500, and display 5 of them
> viewsWithSchemaDF.filter('viewCount>500).limit(5).show
Same query as above with domain of "en" - note the triple = for equality
>viewsWithSchemaDF.filter('domain==="en").filter('viewCount>500).limit
(5).show
```

#### Find the 5 rows with the largest viewCount and a domain of "en"

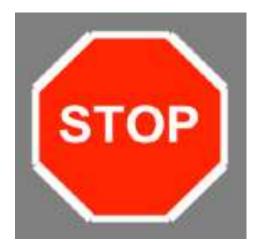
>viewsWithSchemaDF.filter('domain==="en").orderBy('viewcount.desc).li mit(5).show

Find the 5 rows with the largest viewCount and a domain of "en", and where the pageName doesn't contain a colon (":")

>viewsWithSchemaDF.filter('domain==="en").filter(!'pageName.contains(":")).orderBy('viewcount.desc).limit(5).show

#### Summary

We can see that text data can require a little more work than data like JSON with a pre-existing structure. Once you've restructured it with a clear schema, which is not usually difficult, then the full power of DataFrames can be easily applied



# Exercise 9: Dataframe, Datasets and Spark SQL Exploring Grouping

#### Overview

In this lab, we will work with the Github archive that contains activity for a single day. We'll analyze it to find the top contributors for that day.

#### Builds on

None - but you must have the spark shell running.

#### Run time

25-35 minutes

#### Github Activity Archive

You've already worked with the data file (*spark-labs/data/github.json*) that contains a log of all github activity for one day. In previous labs you should already have

Loaded the data, and viewed the schema.

Selected the actor column, which has a nested structure, and worked with some of its subelements.

We illustrate doing that below.

- // Load the file into Spark
- > val githubDF=spark.read.json("spark-labs/data/github.json")
- // Select the actor column
- > val actorDF = githubDF.select('actor)
- // Print actor schema
- > actorDF.printSchema
- // Select the actor's login value note how we
- // Use a SQL expression in the select, not a Column
- > actorDF.select("actor.login").limit(5).show

# Assignment: Query the Data by Actor's Login Value

#### Tasks

➤ Query the github data for how many entries exist in the data for each actor's login. Use the DSL.

#### Hints:

- You'll want to group the data by the actor's login.
- You'll probably want to use an SQL expression to express the actor's login, not a column value.
- You want a count of entries for each login value. Show a few rows of this data.
- ➤ Lastly, find the 20 logins with the largest number of contributions, and display them.

# Assignment 2: Use SQL Tasks

- ➤ Optionally, try doing the above query in SQL.
- ➤ It's pretty much standard SQL, so if you know that well, it's not very complex.
- > Remember to create a temporary view (createOrReplaceTempView)

#### Summary

The task we did in this lab is not overly complex, but it's also not trivial. Spark SQL makes it reasonable for us to accomplish this in a short lab, either using the DSL, or using SQL.

If you wanted to do this using RDDs, it would be a much more complex series of transformations - starting with the messy ones of parsing the JSON data. This is why Spark SQL is becoming the standard API for working with Spark.



# Exercise 10: Seeing Catalyst at Work

#### Overview

In this lab, we will look at several transformations and examine the optimizations that Catalyst performs.

#### Builds on

Previous labs for the transformations we'll use.

#### Run time

20-30 minutes

#### Lab Preparation

We'll first work with the Wikimedia view data, and see how Catalyst helps to optimize queries involving filtering.

#### Tasks

You've already worked with the data file (*spark-labs/data/wiki-pageviews.txt*). In previous labs you should already have done the below. If you haven't, then do it now.

- ➤ Loaded the data, and then split on whitespace.
- Create a new dataframe with a finer-grained schema.
- > We illustrate doing that below.

```
// Load the file into Spark
> val viewsDF=spark.read.text("spark-labs/data/wiki-pageviews.txt")

// Split on whitespace
> val splitViewsDF = viewsDF.select(split('value, "\\s+").as("splitLine"))

// Use a better schema
> val viewsWithSchemaDF = splitViewsDF.select('splitLine(0).as("domain"), 'splitLine(1).as("pageName"), 'splitLine(2).cast("integer").as("viewCount"), 'splitLine(3).cast("long").as("size"))
```

#### # Python

> from pyspark.sql.functions import \*

```
# Load the file into Spark

> viewsDF=spark.read.text("spark-labs/data/wiki-pageviews.txt")

# Split on whitespace

> splitViewsDF = viewsDF.select(split(viewsDF.value, "\s+").alias("splitLine"))

# Use a better schema

> viewsWithSchemaDF = splitViewsDF.select(splitViewsDF.splitLine[0].alias("domain"),splitViewsDF.splitLine[1].alias("pageName"),splitViewsDF.splitLine[2].cast("integer ").alias("viewCount"), splitViewsDF.splitLine[3].cast("long").alias("size"))
```

#### Push Down Predicate

#### Tasks

- First, write a transformation to order viewsWithSchemaDF by viewCount.
- explain the above transformation.
  - Note that there is shuffling involved (an Exchange).
  - It needs to shuffle data to sort the rows.
- ➤ Next, filter viewsWithSchemaDF so you only view rows where the domain starts with "en".
  - Put the filter at the end of the transformation chain in your code (after the ordering).
  - You can use the following on the domain column to accomplish this.
    - o Scala: startsWith("en") (Uppercase W)[5]
    - o **Python:** startswith("en") (Lowercase w)
    - o explain this transformation.
    - You should see the filtering happening before the shuffle for ordering.
    - o Catalyst has **pushed the filter down** to improve efficiency.
    - View the steps that Catalyst took with the detailed version of explain()
      - Scala: explain(true)
      - Python: explain(True) | SEP
- > Optionally, try the transformation with the filter placed before the ordering.
  - It should give you exactly the same plan.

### Scala Only: Work with Datasets and lambdas

We'll now create a Dataset and filter it using a lambda. We'll look at how the lambda affects the Catalyst optimizations.

#### **Tasks**

- ➤ Declare a case class for the Wikimedia data.
- > Create a Dataset using the case class.

We show this code below

- > case class WikiViews(domain:String, pageName:String, viewCount:Integer, size:Long)
- > val viewsDS = viewsWithSchemaDF.as[WikiViews]
  - ➤ Next, filter viewsDS so you only view rows where the domain starts with "en".
    - Put the filter at the end of the transformation chain (after the ordering)
    - You have to use a lambda for this.
    - You can use startsWith("en") on the domain value to accomplish this (this is a Scala function on strings).
    - explain this transformation.

viewsDS.orderBy('viewCount).filter(view=>view.domain.startsWith("en")).explain

➤ Where does the filtering happen? It should be the very last thing. Why?

#### Summary Catalyst rules!

We've explained it, now you see it. But lambdas can overthrow the ruler - so be cautious with them.



## Exercise 11: Joins and Broadcast

#### Overview

In this lab, we will explore some slightly more complex queries, including possible alternatives that use a join.

We'll also look at Catalyst's broadcast optimizations when doing joins, and examine performance with those optimizations in place and without them.

#### Builds on

None

#### Run time

30-40 minutes

#### Top 20 Contributors - Straightforward Query

We want to find out the 20 contributors who have the most entries in the github data. Let's first do this in a straightforward way, as done in an earlier lab.

#### **Tasks**

> Find the top 20 contributors as follows.

```
// Scala
```

> val scanQuery =

githubDF.groupBy("actor.login").count.orderBy('count.desc).limit(20)

> scanQuery.show

#### #Python

```
> scanQuery =
```

githubDF.groupBy("actor.login").count().orderBy("count",ascending=False).limit(20)

> scanQuery.show()

- > Open the Web UI (localhost:4040) and look at the jobs tab.
- Note the execution time (0.6s on our system) | SEP|
- > Drill down on this job, and note the shuffle data (333KB on our system)

## Top 20 Contributors - Join Query

We've decided that we want to track 20 specific contirbutors (the top 20 contributors from the previous month), instead of the top 20 contributors at any given moment. We keep the login values of these contributors-of-interest in another data file we supply (*spark-labs/data/github-top20.json*)

Write a join query that gives the actor login and count from the entries in *github.json* for the login values in *github-top20.json*.

#### Tasks(Assignment)

- Review the data in *github-top20.json*.
- ➤ Accomplish our query needs by:
  - Loading *github-top20.json* into a dataframe (githubTop20DF)
  - Joining githubDF and githubTop20DF in topContributorsJoinedDF
  - Grouping by actor.login
  - Counting the results, and ordering by descending count.

#### Join Query Performance

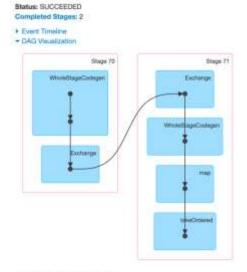
Let's review the performance of this join query, and the optimizations that are being done by Catalyst.

#### **Tasks**

- ➤ Open a browser window on the Web UI localhost:4040.
  - Note the time taken for the last show performed above, drill through on the job for it, and note the shu#e data.
  - On our system it took 0.6s and shuffled 7.9K of data.
  - This is much less data shuffling than the scanQuery done earlier (40 times reduction).
- > Execute topContributorsJoinedDF.explain to see what is going on.
  - We illustrate this below. Note the broadcasts which we have highlighted.
  - Catalyst has optimized the join to use broadcast.

```
lettlevesque!
       LukasReschkel
                       281
        nwt-patrick|
                       271
            Somasis
                       27
         mikegazdag|
                       261
        tterrag1098|
                       231
    EmanueleMinotto|
scala> topContributorsJoinedDF.explain()
== Physical Plan ==
*(4) Sort [count#540L DESC NULLS LAST], true, 0
+- Exchange rangepartitioning(count#540L DESC NULLS LAST, 200)
   +- *(3) HashAggregate(keys=[actor#6.login#556], functions=[count(1)])
      +- Exchange hashpartitioning(actor#6.login#556, 200)
         +- *(2) HashAggregate(keys=[actor#6.login AS actor#6.login#556], functions=[partial_count(1)])
            +- *(2) Project [actor#6]
               +- *(2) BroadcastHashJoin [actor#6.login], [login#421], Inner, BuildRight
                  :- *(2) Project [actor#6]
                  : +- *(2) Filter isnotnull(actor#6)
                         +- *(2) FileScan json [actor#6] Batched: false, Format: JSON, Location: InMemor
yFileIndex[file:/Users/neha/Desktop/Neha/Spark/spark-labs/data/github.json], PartitionFilters: [], Push
edFilters: [IsNotNull(actor)], ReadSchema: struct<actor:struct<avatar_url:string,gravatar_id:string,id:
bigint, login: string, url: string>>
                  +- BroadcastExchange HashedRelationBroadcastMode(List(input[0, string, true]))
                      +- *(1) Project [login#421]
                        +- *(1) Filter isnotnull(login#421)
                            +- *(1) FileScan json [login#421] Batched: false, Format: JSON, Location: In
MemoryFileIndex[file:/Users/neha/Desktop/Neha/Spark/spark-labs/data/github-top20.json], PartitionFilter
s: [], PushedFilters: [IsNotNull(login)], ReadSchema: struct<login:string>
scala>
```

#### Details for Job 36



#### - Completed Stages (2)

Stage Id +	Description		Submitted	Duration	Tasks: Succeeded/Total	Input	Output	Shuffle Read	Shuffle Write
71	show at -consoler-32	+details	2020/06/03 00:41:46	87 ms	200/200			9.6 KB	
70	show at <ponsole>:32</ponsole>	+details	2020/06/03 00:41:46	67 ms	11/11	43.6 MB			9.6 KB

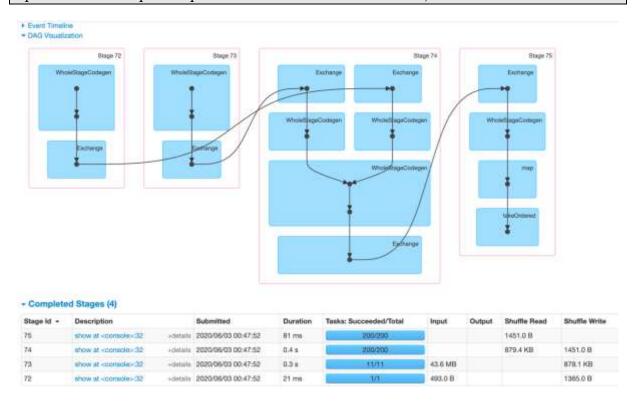
### Join Performance without Broadcast

Let's disable the Catalyst broadcast optimization and look at what happens.

#### Tasks:

- ➤ Disable automatic broadcasting, by seting the appropriate configuration property as shown below.
- > Next, create and show the topContributorsJoinedDF query again
  - SEP You must create the DataFrame after setting the config property, to make sure it conforms to the new setting.

## spark.conf.set("spark.sql.autoBroadcastJoinThreshold",-1)



```
|scala> topContributorsJoinedDF.explain()
== Physical Plan ==
*(7) Sort [count#540L DESC NULLS LAST], true, 0
+- Exchange rangepartitioning(count#540L DESC NULLS LAST, 200)
   +- *(6) HashAggregate(keys=[actor#6.login#568], functions=[count(1)])
      +- Exchange hashpartitioning(actor#6.login#568, 200)
         +- *(5) HashAggregate(keys=[actor#6.login AS actor#6.login#568], functions=[partial_count(1)])
             +- *(5) Project [actor#6]
               +- *(5) SortMergeJoin [actor#6.login], [login#421], Inner
                  :- *(2) Sort [actor#6.login ASC NULLS FIRST], false, 0
                     +- Exchange hashpartitioning(actor#6.login, 200)
                        +- *(1) Project [actor#6]

← *(1) Filter isnotnull(actor#6)
                               +- *(1) FileScan json [actor#6] Batched: false, Format: JSON, Location: I
nMemoryFileIndex[file:/Users/neha/Desktop/Neha/Spark/spark-labs/data/github.json], PartitionFilters: []
, PushedFilters: [IsNotNull(actor)], ReadSchema: struct<actor:struct<avatar_url:string,gravatar_id:stri
ng,id:bigint,login:string,url:string>>
                  +- *(4) Sort [login#421 ASC NULLS FIRST], false, 0
                     +- Exchange hashpartitioning(login#421, 200)
                        +- *(3) Project [login#421]
                           +- *(3) Filter isnotnull(login#421)
                              +- *(3) FileScan json [login#421] Batched: false, Format: JSON, Location:
 InMemoryFileIndex[file:/Users/neha/Desktop/Neha/Spark/spark-labs/data/github-top20.json], PartitionFil
ters: [], PushedFilters: [IsNotNull(login)], ReadSchema: struct<login:string>
scala>
```

- ➤ Observe the execution time and shuffled data. This is almost **100 times** the amount of shuffle data!
- ➤ Renable automatic broadcasting, by seting the appropriate configuration property as shown below.

> spark.conf.set("spark.sql.autoBroadcastJoinThreshold",1024\*1024\*10)

#### Summary

Joins are powerful tools, but distributed joins can be expensive in terms of amount of data shu#ed. We've worked with them here, and seen how Catalyst can optimize by broadcasting a small data set.

Note that we've used a common strategy here - we've replaced a query that was resource intensive (our straight scan query), with a di!erent query (querying on a specific group of logins that are provided). The new query is not exactly equivalent, but fills our needs, and is much less resource intensive.



## Exercise 12: Hive with Spark integration



Let's create table "reports" in the hive. I am using boa schema in which I am creating a table.

Enter in to hive CLI and use below commands to create a table:

CREATE schema boa;

CREATE TABLE boa.reports(id INT,days INT,YEAR INT);

INSERT INTO TABLE boa.reports VALUES (121,232,2015),(122,245,2015),(123,134,2014),(126,67,2016),(182,122,2016),(137,92,2015),(101,311,2015);

select \* from boa.reports;

➤ Go to spark-shell:

Now, create a data frame hiveReports using below command:

var hiveReports = spark.sqlContext.sql("select \* from boa.reports")

> Check whether dataset report is loaded into data frame hiveReport or not using below command:

Check schema of Data Frame.

➤ hiveReports.show() will show the same results as "select \* from boa.reports" in Hive CLI.

Accessing Catalog from spark-shell Catalog is available on spark session. The following code shows how to access catalog.

#### val catalog = sparkSession.catalog

Querying the databases Once we have access to catalog, we can use it to query the databases. All the API's on catalog returns a dataset.

#### catalog.listDatabases().select("name").show()

Registering Dataframe with createTempView
In earlier versions of spark, we used to register a dataframe using
registerTempTable. But in spark 2.0, this API is deprecated. The
registerTempleTable API was one of the source of confusion as users used think
it materializes the dataframe and saves as a temporary table which was not the
case. So this API is replaced with createTempView.

createTempView can be used as follows.

## df.createTempView("sales")

spark.catalog.listTables("boa").show()

Once we have registered a view, we can query it using listTables.

Checking is table cached or not Catalog not only is used for querying. It can be used to check state of individual tables. Given a table, we can check is it cache or not. It's useful in scenarios to make sure we cache the tables which are accessed frequently.

#### catalog.isCached("sales")

You will get false as by default no table will be cache. Now we cache the table and query again.

#### df.cache()

catalog.isCached("sales")

Now it will print true.

#### > Drop view

We can use catalog to drop views. In spark sql case, it will deregister the view. In case of hive, it will drops from the metadata store.

#### catalog.dropTempView("sales")

Query registered functions

Catalog API not only allow us to interact with tables, it also allows us to interact with udf's. The below code shows how to query all functions registered on spark session. They also include all built in functions.

catalog.listFunctions (). select ("name", "description", "className", "isTemporary"). s how (100)



## Exercise 13: SQL Tables and Views

#### CREATING A MANAGED TABLE

Tables reside within a database. By default, Spark creates tables under the default database. To create your own database name, you can issue a SQL command from your Spark application or notebook. Using the US flight delays data set, let's create both a managed and an unmanaged table. To begin, we'll create a database called training\_spark\_db and tell Spark we want to use that database:

// In Scala/Python

Launch

#### bin/spark-sql -S

for spark-sql console

#### CREATE DATABASE learn\_spark\_db

```
spark-sql> CREATE DATABASE learn_spark_db;
spark-sql> show databases;
bdp
default
demohivespark
learn_spark db
spark-sql> 

### April 19  

### April 20  

##
```

#### Launch bin/spark-shell

# spark.catalog.listDatabases.show(false) scala> spark.catalog.listDatabases.show(false) 20/07/05 11:08:19 WARN DomainSocketFactory: The short-circuit local reads feature cannot be used because libhadoop cannot be loaded. |locationUri Inate |description [hdfs://duickstart.cloudera:8020/user/hive/warehouse/bdp.db 1bdp default hdfs://quickstart.cloudera.8020/user/hive/warehouse hdfs://quickstart.cloudera:8020/user/hive/warehouse/demohivespark.db demohivespark learn\_spark\_db hdfs://quickstart.cloudera:8020/user/hive/warehouse/learn\_spark\_db.db spark.sql("USE training\_spark\_db") spark.sql("CREATE TABLE managed\_us\_delay\_flights\_tbl (date STRING, delay INT, distance INT, origin STRING, destination STRING)") val csvFile="file:///home/cloudera/Desktop/SW/in/Data/departuredelays.csv" val schema=date STRING, delay INT, distance INT, origin STRING, destination STRING" val df = spark.read.format("csv").option("schema", schema).option("header", "true").load(csvFile) import org.apache.spark.sql.{Row, SaveMode} df.write.mode(SaveMode.Overwrite).saveAsTable("managed us delay flights tbl ")

#### **CREATING AN UNMANAGED TABLE**

By contrast, you can create unmanaged tables from your own data sources—say, Parquet, CSV, or JSON files stored in a file store accessible to your Spark application.

To create an unmanaged table from a data source such as a CSV file, in SQL use:

```
spark.sql("""CREATE TABLE us_delay_flights_tbl(date STRING, delay INT,
 distance INT, origin STRING, destination STRING)
 USING csv OPTIONS (PATH
 'file:///home/cloudera/Desktop/SW/in/Data/departuredelays.csv')""")
df.write.mode(SaveMode.Overwrite).option("path",
"/tmp/data/us_flights_delay").saveAsTable("us_delay_flights_tbl")
scala> df.write.mode(SaveMode.Overwrite).option("path", "/tmp/data/us flights delay").saveAsTable("us delay flights tbl")
scala> spark.sql("select * from us_delay_flights_tbl limit 5").show()
    date|delay|distance|origin|destination|
01011245
010206001
                3691
                     ABE
                               DTW
01021245
                662
                      ABE
                               ATL
 010206051
010312451
                      ABEI
                               ATL
```

# Exercise 14: Hive with Spark integration using IntelliJ

Update pom.xml with the below dependency for hive libraries

```
<dependency>
<groupId>org.apache.spark</groupId>
```

```
<artifactId>spark-hive_${scala.tools.version}</artifactId>
<version>${spark.version}</version>
</dependency>
```

Let us create a Scala Object in the existing workspace of training-spark, with the below provided code.

```
import java.io.File
import org.apache.spark.sql.{Row, SaveMode, SparkSession}
case class Record(key: Int, value: String)
// warehouseLocation points to the default location for managed databases and
val warehouseLocation = new File("spark-warehouse").getAbsolutePath
val spark = SparkSession
 .builder()
 .appName("Spark Hive Example") .master("local[*]")
//.config("spark.sql.warehouse.dir", warehouseLocation)
 .enableHiveSupport()
 .getOrCreate()
import spark.implicits._
import spark.sql
sql("CREATE TABLE IF NOT EXISTS src (key INT, value STRING) USING
hive")
sql("LOAD DATA LOCAL INPATH 'examples/src/main/resources/kv1.txt' INTO
TABLE src")
// Queries are expressed in HiveQL
sql("SELECT * FROM src").show()
// +---+
```

```
// |key| value|
// +---+
// |238|val_238|
// | 86 | val_86 |
// |311|val_311|
// ...
// Aggregation queries are also supported.
sql("SELECT COUNT(*) FROM src").show()
// +----+
// | count(1) |
// +----+
// | 500 |
// +----+
// The results of SQL queries are themselves DataFrames and support all normal
functions.
val sqlDF = sql("SELECT key, value FROM src WHERE key < 10 ORDER BY
key")
// The items in DataFrames are of type Row, which allows you to access each
column by ordinal.
val stringsDS = sqlDF.map {
 case Row(key: Int, value: String) => s"Key: $key, Value: $value"
stringsDS.show()
// +----+
// |
           value
// +----+
// | Key: 0, Value: val_0 |
// | Key: 0, Value: val_0 |
// | Key: 0, Value: val_0 |
```

```
// ...
// You can also use DataFrames to create temporary views within a
SparkSession.
val recordsDF = spark.createDataFrame((1 to 100).map(i => Record(i, s"val_$i")))
recordsDF.createOrReplaceTempView("records")
// Queries can then join DataFrame data with data stored in Hive.
sql("SELECT * FROM records r JOIN src s ON r.key = s.key").show()
// +---+
// |key | value | key | value |
// +---+
// | 2 | val_2 | 2 | val_2 |
// | 4 | val_4 | 4 | val_4 |
// | 5 | val_5 | 5 | val_5 |
// ...
// Create a Hive managed Parquet table, with HQL syntax instead of the Spark
SQL native syntax
// `USING hive`
sql("CREATE TABLE hive_records(key int, value string) STORED AS
PARQUET")
// Save DataFrame to the Hive managed table
val df = spark.table("src")
df.write.mode(SaveMode.Overwrite).saveAsTable("hive_records")
// After insertion, the Hive managed table has data now
sql("SELECT * FROM hive_records").show()
// +---+
// |key| value|
// +---+
// |238|val 238|
// | 86 | val_86 |
```

```
// |311|val_311|
// ...
// Prepare a Parquet data directory
val dataDir = "/tmp/parquet_data"
spark.range(10).write.parquet(dataDir)
// Create a Hive external Parquet table
sql(s"CREATE EXTERNAL TABLE hive_bigints(id bigint) STORED AS
PARQUET LOCATION '$dataDir'")
// The Hive external table should already have data
sql("SELECT * FROM hive_bigints").show()
// +---+
// | id|
// +---+
// | 0|
// | 1|
// | 2|
// ... Order may vary, as spark processes the partitions in parallel.
// Turn on flag for Hive Dynamic Partitioning
spark.sqlContext.setConf("hive.exec.dynamic.partition", "true")
spark.sqlContext.setConf("hive.exec.dynamic.partition.mode", "nonstrict")
// Create a Hive partitioned table using DataFrame API
df.write.partitionBy("key").format("hive").saveAsTable("hive_part_tbl")
// Partitioned column `key` will be moved to the end of the schema.
sql("SELECT * FROM hive_part_tbl").show()
// +----+
// | value | key |
// +----+
// |val_238|238|
// | val_86 | 86 |
```

#### // |val\_311|311|

#### spark.stop()

-master yarn

Clean and build the jar using maven tools and submit t he spark-submit command

bin/spark-submit --class com.sparkTutorial.hiveintegration.HiveIntegrationSpark /home/cloudera/Desktop/SW/Code/Lesson4/Lesson4/target/trainingSpark-1.0.jar -

Observe the output from the console and hive table created from CLI.

The metastore in Hive will store, as below

```
(base) [cloudera@quickstart spark]$ hdfs dfs -ls /user/hive/warehouse
Found 7 items
drwxrwxrwx

    cloudera supergroup

                                             0 2020-06-03 06:03 /user/hive/warehouse/bdp.db
            - cloudera supergroup
                                             0 2020-06-04 01:03 /user/hive/warehouse/demohivespark.db
drwxrwxrwx
drwxrwxrwx

    cloudera supergroup

                                             0 2020-06-05 23:51 /user/hive/warehouse/hive part tbl
            - cloudera supergroup
                                             0 2020-06-05 23:50 /user/hive/warehouse/hive_records
drwxr-xr-x
                                             0 2020-06-05 04:55 /user/hive/warehouse/partītioned user
drwxrwxrwx

    cloudera supergroup

            - cloudera supergroup
- cloudera supergroup
                                             0 2020-06-05 23:49 /user/hive/warehouse/src
drwxrwxrwx
                                             0 2020-06-05 04:48 /user/hive/warehouse/users_part
drwx rwx rwx
(base) [cloudera@quickstart spark]$ |
```



# Exercise 15: Produce and Consume Apache Kafka Messages

# Start the Zookeeper server

1. Open a new terminal window and browse to the location "/usr/lib/zookeeper" using cd. Start the zookeeper by command Navigate to /home/cloudera/Desktop/SW/zookeeper-3.4.14.

bin/zkServer.sh start

```
[cloudera@quickstart ~]$ cd /usr/lib/zookeeper
[cloudera@quickstart zookeeper]$ bin/zkServer.sh start
JMX enabled by default
Using config: /usr/lib/zookeeper/bin/../conf/zoo.cfg
Starting zookeeper ... STARTED
[cloudera@quickstart zookeeper]$
```

2. You can now validate that Zookeeper is running correctly in standalone mode by connecting to the client port and sending the four letter command "srvr".

```
telnet localhost 2181
```

```
[cloudera@quickstart zookeeper]$ telnet localhost 2181
Trying 127.0.0.1...
Connected to localhost.
Escape character is '^]'.
srvr
Zookeeper version: 3.4.5-cdh5.12.0--1, built on 06/29/2017 11:30 GMT
Latency min/avg/max: 0/0/1303
Received: 30491
Sent: 30604
Connections: 4
Outstanding: 0
Zxid: 0xe14
Mode: standalone
Node count: 737
Connection closed by foreign host.
[cloudera@quickstart zookeeper]$ ■
```

## Starting Kafka and Creating a Kafka Topic

1. Open a new terminal and start Kafka broker, Change the user to cloudera with su - - and provide password as cloudera.

Navigate to /home/cloudera/Desktop/SW/kafka\_2.11-2.2.1

bin/kafka-server-start.sh config/server.properties

```
[2018-83-11 80:04:10,389] IMFO starting (Nafka.server.KafkaServer)
[2018-83-11 80:04:10,480] IMFO Connecting to Joukeper on localhostizIR1 (Nafka.server.KafkaServer)
[2018-83-10 80:04:10,490] IMFO Connecting to Joukeper on localhostizIR1 (Nafka.server.KafkaServer)
[2018-83-10 80:04:10,490] IMFO Claim environment incohesper.version=3.4,5-of5.9.6-1, built on 10/21/2016 80:85 GMT (org.apache.zookeeper.Zookeeper.
[2018-83-10 80:04:10,490] IMFO Claim environment incohesper.version=3.4,5-of5.9.6-1, built on 10/21/2016 80:85 GMT (org.apache.zookeeper.Zookeeper.
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.Zookeeper.
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.Zookeeper.
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.)
[2018-83-11 80:04:10,490] IMFO Claim environment java.version=1.7.6 87 (org.apache.zookeeper.Zookeeper.)
```

Kafka broker is running on port 9042.

2. Open a new terminal window and create a Kafka topic named weblogs that will contain messages representing lines in Loudacre's web server logs.

Since your exercise environment is a single-node cluster running on a virtual machine, use a replication factor of 1 and a single partition.

Navigate to the location where Kafka is installed, which is "/usr/lib/kafka".

```
$ bin/kafka-topics.sh --create \ --zookeeper localhost:2181 \ --
replication-factor 1 \ --partitions 1 \
--topic weblogs
```

Here, Topic name

The name of the topic that you wish to create.

Replication Factor

The number of replicas of the topic to maintain within the cluster.

**Partitions** 

The number of partitions to create for the topic.

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh --create --zookeeper localhost:ZIB1 --replication-factor 1 --partitions 1 --topic weblogs SLF41: Class path contains multiple SLF41 bindings.

SLF41: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4]12-1.7.Z1.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF41: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4]12-1.7.Z1.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF41: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4]12-1.7.Z.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF41: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]

Created topic "weblogs".
```

1. Display all Kafka topics to confirm that the new topic you just created is listed:

```
$ kafka-topics --list --zookeeper localhost:2181
```

```
[cloudera@quickstart kafka]$ kafka-topics --list --zookeeper localhost:2181

SLF4J: Class path contains multiple SLF4J bindings.

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.21.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.5.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF4J: See http://www.slf4j.org/codes.html#multiple bindings for an explanation.

SLF4J: Actual binding is of type [org.slf4j.impl.Log4jloggerFactory]

weblogs
[cloudera@quickstart kafka]$ [
```

2. Describe the Kafka topic created

```
$ bin/kafka-topics.sh --describe --zookeeper localhost:2181
--topic weblogs
```

# Producing and Consuming Messages

You will now use Kafka command line utilities to start producers and consumers for the topic created earlier.

3. Start a Kafka producer for the weblogs topic:

```
$ bin/kafka-console-producer.sh --broker-list localhost:9092 --topic
weblogs
```

**Tip**: This exercise involves using multiple terminal windows. To avoid confusion, set a different title for each one by selecting Set Title... on the Terminal menu:

Set the title for this window to "Kafka Producer."

4. Publish a test message to the weblogs topic by typing the message text and then pressing Enter. For example:

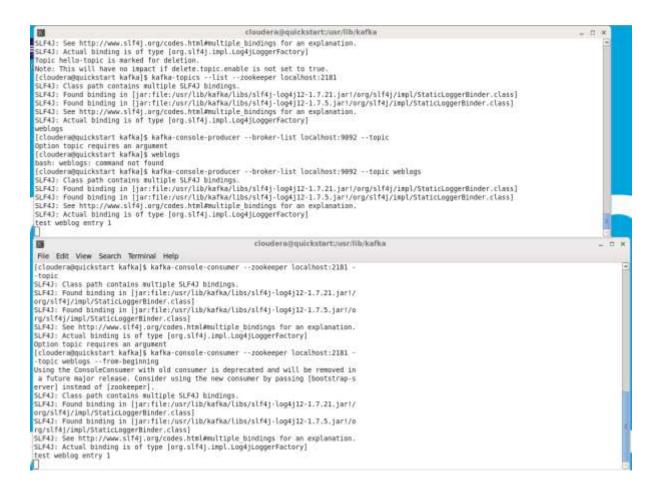
```
test weblog entry 1
```

- 5. Open a new terminal window and adjust it to fit on the window beneath the producer window. Set the title for this window to "Kafka Consumer."
- 6. In the new terminal window, start a Kafka consumer that will read from the beginning of the weblogs topic:

```
$ bin/kafka-console-consumer.sh --bootstrap-server localhost:9092 --
topic
weblogs --from-beginning
```

You should see the status message you sent using the producer displayed on the consumer's console, such as:

test weblog 1



- 7. Press Ctrl+C to stop the weblogs consumer, and restart it, but this time omit the from-beginning option to this command. You should see that no messages are displayed.
- 8. Switch back to the producer window and type another test message into the terminal, followed by the Enter key:

```
test weblog entry 2
```



9. Return to the consumer window and verify that it now displays the alert message you published from the producer in the previous step.

# Cleaning Up

- 10. Press Ctrl+C in the consumer terminal window to end its process.
- 11. Press Ctrl+C in the producer terminal window to end its process.



# Exercise 16: Alter Apache Kafka Topics

Let us try altering some parameters for the topic, but instead of using the earlier created topic <weblogs>, we will create another topic "hello-topic" to alter some parameters.

Exercise: Create a topic called hello-topic with replication factor 1 and partition 1.

1. We will add more partitions in the topic created as "hello-topic". Below command will add 10 more partitions to the hello-topic topic. Note that before, the topic has only 1 partition.

```
bin/kafka-topics.sh --alter --zookeeper localhost:2181 --
partitions 11 --topic hello-topic
```

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh --alter --zookeeper localhost:2181 --partitions 11 --topic hello-topic SLF4J: Class path contains multiple SLF4J bindings.

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.21.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.5.jar!/org/slf4j/impl/StaticLoggerBinder.class]

SLF4J: See http://www.slf4j.org/codes.html#multiple bindings for an explanation.

SLF4J: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]

WARNING: If partitions are increased for a topic that has a key, the partition logic or ordering of the messages will be affected 
Adding partitions succeeded!

[cloudera@quickstart kafka]$
```

Let us verify the number of partitions is being changed with the command

```
bin/kafka-topics.sh --describe --zookeeper localhost:2181 --
topic hello-topic
```

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh --describe --zookeeper localhost:2181 --topic hello-topic
SLF4J: Class path contains multiple SLF4J bindings.
SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.21.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4): Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4j12-1.7.5.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: See http://www.slf4j.org/codes.html#multiple bindings for an explanation.
SLF4J: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]
                           PartitionCount:11
Topic:hello-topic
                                                      ReplicationFactor:
         Topic: hello-topic
Topic: hello-topic
                                    Partition: 0
                                                      Leader: 0
                                                                        Replicas: 0
                                                                                           Isr: 0
                                    Partition: 1
                                                      Leader: 8
                                                                        Replicas: 0
                                                                                           Isr: 8
         Topic: hello-topic
                                    Partition: 2
                                                      Leader: 0
                                                                        Replicas: 0
                                                                                           Isr: 0
         Topic: hello-topic
                                    Partition: 3
                                                      Leader: 0
                                                                        Replicas: 0
                                                                                           Isr: 0
         Topic: hello-topic
                                    Partition: 4
                                                      Leader: 0
                                                                        Replicas: 6
                                                                                           Isr: 0
                                                      Leader: 0
                                                                        Replicas: 0
                                                                                           Isr: 0
         Topic: hello-topic
                                    Partition: 5
         Topic: hello-topic
                                    Partition: 6
                                                      Leader: 0
                                                                        Replicas: 0
                                                      Leader: 0
Leader: 0
         Topic: hello-topic
                                    Partition: 7
                                                                        Replicas: 0
                                                                                           Isr: 0
                                    Partition: B
                                                                        Replicas: 0
                                                                                           Isr: 0
         Topic: hello-topic
         Topic: hello-topic
                                                      Leader: 0
                                                                        Replicas: 0
                                    Partition: 9
         Topic: hello-topic
                                    Partition: 10
                                                      Leader: 0
                                                                        Replicas: 0
                                                                                           Isr: 0
[cloudera@quickstart kafka]s
```

#### Tip: REDUCING PARTITION COUNTS

It is not possible to reduce the number of partitions for a topic. The reason this is not supported is because deleting a partition from a topic would cause part of the data in that topic to be deleted as well, which would be inconsistent from a client point of view. In addition, trying to redistribute the data to remaining partitions would be difficult and result in out-of-order messages. Should you need to reduce the number of partitions, you will need to delete the topic and recreate it.

#### 2. Add configurations to the Kafka topic

The general syntax is:

```
bin/kafka-topics.sh --alter --zookeeper localhost:2181 --topic
kafkatopic --config <key>=<value>
```

There are various configuration fields we can set for the topic, here we will set up the max.message.bytes - this is the largest size of the message the broker will allow to be appended to the topic. This size is validated pre-compression. (Defaults to broker's message.max.bytes.)

```
bin/kafka-topics.sh --zookeeper localhost:2181 --alter --topic
hello-topic --config max.message.bytes=128000
```

```
[cloudera@quickstart kafka]s bin/kafka-topics.sh --zookeeper localhost:2181 --alter --topic hello-topic --config max.message.bytes=128000 SLF4J: Class path contains multiple SLF4J bindings.

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4j12-1.7.21.jarl/org/slf4]/impl/StaticLoggerBinder.class|

SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4j12-1.7.5.jarl/org/slf4]/impl/StaticLoggerBinder.class|

SLF4J: Goo http://www.slf4j.org/codes.html#multiple bindings for an explanation.

SLF4J: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]

WARMING: Altering topic configuration from this script has been deprecated and may be removed in future releases.

Going forward, please use kafka-configs.sh for this functionality

Updated config for topic "hello-topic".
```

3. To remove above overridden configuration, we can use command:

bin/kafka-topics.sh --zookeeper localhost:2181 --alter --topic
hello-topic --delete-config max.message.bytes



## Exercise 18: Delete a topic

If a topic is no longer needed, it can be deleted in order to free up these resources. In order to perform this action, the brokers in the cluster must have been configured with the delete.topic.enable option set to true. If this option has been set to false, then the request to delete the topic will be ignored.

We will delete the topic created with the below command

```
bin/kafka-topics.sh --delete --zookeeper localhost:2181 --
topic hello-topic
```

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh --delete --zookeeper localhost:2181 --topic hello-topic
5LF4J: Class path contains multiple SLF4J bindings.
5LF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.21.jar!/org/slf4j/impl/StaticLoggerBinder.class]
5LF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.5.jar!/org/slf4j/impl/StaticLoggerBinder.class]
5LF4J: See http://www.slf4j.org/codes.html#multiple bindings for an explanation.
5LF4J: Actual binding is of type [org.slf4j.impl.Log4jLoggerFactory]
Topic hello-topic is marked for deletion.
Note: This will have no impact if delete.topic.enable is not set to true.
```

Lets verify if the topic has been deleted, with the -list command

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh --zookeeper localhost:2181 --list
SLF4J: Class path contains multiple SLF4J bindings.
SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.21.jarl/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: Found binding in [jar:file:/usr/lib/kafka/libs/slf4j-log4j12-1.7.5.jarl/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: See http://www.slf4j.org/codes.html#multiple_bindings for an explanation.
SLF4J: Actual binding is of type [org.slf4j.impl.Log4jloggerFactory]
weblogs
[cloudera@quickstart kafka]$ |
```

#### Tip: DATA LOSS AHEAD

Deleting a topic will also delete all its messages. This is not a reversible operation, so make sure it executed carefully.



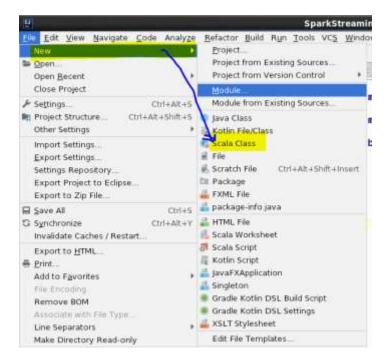
# Exercise 19: Building Spark Application using IntelliJ IDE Apache Kafka Scala Client API

Goal: We are going to create a producer and consumer by using **Apache Kafka Scala** client **API**.

1. Create new SBT Project using the menu bar option File ->New Project -> Select Scala and SBT. Give a name of your choice for the project.



Under src→main→scala, Add a new Scala class.



2. The content of build.sbt

```
version := "1.0"

scalaVersion := "2.11.6"

libraryDependencies ++= Seq (
"org.apache.kafka" %"kafka-clients" %"0.11.0.0
)
```

Note: A pop up of "Enable Auto Import", please click on it to enable it.

- 3. Create a KafkaProducer Scala Object, as below:-
- 4. Now add a new Scala class file to project, Right click on project > New > ScalaClass > KafkaProducerTest.

```
object KafkaProducerTest extends App {

import Scala.util.Properties

import org.apache.kafka.clients.producer._
```

```
val props = new Properties()
props.put("bootstrap.servers", "localhost:9092")
props.put("acks","1")
props.put("key.serializer", "org.apache.kafka.common.serialization.StringSerializer")
props.put("value.serializer", "org.apache.kafka.common.serialization.StringSerializer")

val producer = new KafkaProducer[String, String](props)

val topic="testKafka"
for(i<- 1 to 50) {
   val record = new ProducerRecord(topic, "key"+i, "value"+i)
   producer.send(record)
   }
   println("sent")
   producer.close()
}</pre>
```

```
SparkStreaming [-]IdeaProjects/SparkStreaming] - .../src/main/scala/KafkaProducer.scala [Sp [Fe [dit View Havigate Code Arwlyze Befactor [Julid Run Tools VC] Window Help

SparkStreaming | More | Moreon |

SparkStreaming | More | Moreon | Moreon | Moreon | Moreon | Moreon |

SparkStreaming | Moreon | Moreon | Moreon | Moreon | Moreon |

SparkStreaming | Julid | Moreon | Moreon | Moreon | Moreon |

SparkStreaming | Julid | Julid | Moreon | Moreon | Moreon |

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SparkStreaming | Julid | Julid | Julid | Julid | Julid |

SparkStreaming | Julid | Julid | Julid | Julid | Julid | Julid |

SparkStreaming | Julid | Julid
```

A Kafka producer has three mandatory properties:

#### bootstrap.servers

List of host:port pairs of brokers that the producer will use to establish initial connection to the Kafka cluster. This list doesn't need to include all brokers, since the producer will get more information after the initial connection. But it is

recommended to include at least two, so in case one broker goes down, the producer will still be able to connect to the cluster.

#### key.serializer

Name of a class that will be used to serialize the keys of the records we will produce to Kafka. Kafka brokers expect byte arrays as keys and values of messages. The producer has to know how to convert these objects to byte arrays. key.serializer should be set to a name of a class that implements the org.apache.kafka.common.serialization.Serializer interface. The producer will use this class to serialize the key object to a byte array.

The Kafka client package includes

ByteArraySerializer (which doesn't do much),

StringSerializer, and

IntegerSerializer, so if you use common types, there is no need to implement your own serializers.

Setting key.serializer is required even if you intend to send only values.

#### value.serializer

Name of a class that will be used to serialize the values of the records we will produce to Kafka. The same way you set key.serializer to a name of a class that will serialize the message key object to a byte array, you set value.serializer to a class that will serialize the message value object.

5. Now let us add a new Scala class file to project, Right click on project >New > Class > KafkaConsumerTest

import scala.collection.ScalaConverters						
import Scala.util						
import Scala.util.Properties						
import org.apache.kafka.clients.consumer.{ConsumerRecords, KafkaConsumer}						
object KafkaConsumerTest extends App {						

```
val properties = new Properties()
properties.put("bootstrap.servers", "localhost:9092")
properties.put("group.id", "KafkaExampleNewConsumer")
properties.put("key.deserializer", "org.apache.kafka.common.serialization.StringDeserializer")
properties.put("value.deserializer", "org.apache.kafka.common.serialization.StringDeserializer")
properties.put("auto.offset.reset", "latest")
val consumer = new KafkaConsumer[String, String](properties)

consumer.subscribe(util.Arrays.asList("testKafka"))

while (true) {
    val records: ConsumerRecords[String, String] = consumer.poll(1000)
    records.asScala.foreach(record => println(s"Received message: "+record))
}
```

```
import scala.collection.JavaConverters
import org.apache.kafka.clients.consumer.
 import org.apache.kafka.common.serialization.StringDeserializer
import java.util.{Properties, UUID}
import java.util
import scala.collection.JavaConverters._
object KafkaConsumer (
   def main(args: Array[String]): Unit = {
     val properties = new Properties()
     properties.put("bootstrap.servers", "localhost:9092")
     properties.put("group.id", "KafkaExampleNewConsumer")
properties.put("key.deserializer", "org.apache.kafka.common.serialization.StringDeserializer")
properties.put("value.deserializer", "org.apache.kafka.common.serialization.StringDeserializer")
properties.put("auto.offset.reset", "latest")
     val consumer = new KafkaConsumer[String, String](properties)
     consumer.subscribe(util.Arrays.asList("testKafka"))
        while (true) {
          val records: ConsumerRecords[String, String] = consumer.poll(timeout = 1000)
          records.asScala.foreach(record => println(s"Received message: $record"))
     }
}
```

6. Now we will test our KafkaProducerTest sending some messages, which will be received by KafkaConsumerTest, before testing ensure that

Pre-requisites:-

Topic 'testKafka' is created.

Zookeeper should be running on localhost:2181.

Kafka should be running on localhost:9092.

#### Run ProducerTest.Scala > Scala Application.

```
Start-- sending messages
Sent ->Messagel
Sent ->Message2
Sent ->Message3
Sent ->Message4
Sent ->Message5
Sent ->Message6
Sent ->Message7
Sent ->Message8
Sent ->Message9
Sent ->Message10
Sent ->Message11
Sent ->Message12
        ->Message13
Sent
Sent ->Message14
Sent ->Message15
Sent ->Message16
Sent ->Message17
       ->Message19
Sent
Sent
Sent ->Message21
Sent ->Message22
Sent ->Message23
Sent ->Message24
Sent ->Message25
Sent ->Message26
       ->Message27
Sent ->Message28
Sent ->Message29
Sent ->Message30
Sent
       ->Message31
Sent ->Message32
Sent ->Message33
5ent ->Message34
```

Right click on the scala class→Run KafkaConsumerTest.scala, the console of the receiver will receive all the messages from the broker. The console will print the key and the message.



Terminate the KafkaConsumerTest.Scala class.

# Assignment: In this assignment, we will use Sending a Message Synchronously

- 1. Create a Producer class as "KafkaPublisher" for the topic "topic-1".
- 2. Imports
  - import org.apache.kafka.clients.consumer.ConsumerRecords;
  - import org.apache.kafka.clients.consumer.ConsumerRecord;
  - import org.apache.kafka.clients.producer.ProducerRecord;
- 3. Set the mandatory values for the KafkaPublisher, which is the producer class bootstrap.servers=broker2:9092
  - key.serializer=org.apache.kafka.common.serialization.StringSerializer
  - value.serializer=org.apache.kafka.common.serialization.StringSerializer
- 4. Set the values in a Properties object.
- 5. Instead of Fire-and-forget which was done before we will use the synchronous way of sending the message, let us send a message, the send() method returns a Future object, and we use get() to wait on the future and see if the send() was successful or not.
- 6. The message record is created as

val record: ProducerRecord[String, String] = new ProducerRecord(TOPIC, inlineMessage)

The ProducerRecord objects we created included a topic name and value.

7. Send the message i.e. record using val futureResponse: Future[RecordMetadata] = producer.send(record) futureResponse.isDone

in the KafkaPublisher.Scala class.

- 8. Create a Consumer class as KafkaSubscriber.
- 9. Set the three mandatory properties: bootstrap.servers, key.deserializer, and value.deserializer (deserializer should be of same type as of serializer).
- 10. Now set the fourth property, which is not strictly mandatory, but for now we will pretend it is. The property is group.id and it specifies the consumer group the KafkaSubscriber instance belongs to.

```
props.put("group.id", "testgroup")
```

- 11. Poll the message on the topic mytopic using the poll() method
- 12. Run the KafkaPublisher as Scala Application and also run the KafkaSubscriber as Scala Application.
- 13. The received message will be printed in the console.

#### **Producer**

```
KafkaSyncConsumer -
                          Kafka5yncProduce
       /usr/java/jdk1.8.0_131/bin/java
      log4j:WARN No appenders could be found for logger (org.apache.kafka.clients.producer.ProducerConfig). log4j:WARN Please initialize the log4j system properly.
11
       log4):WARN See http://logging.apache.org/log4j/1.2/faq.html#moconfig for more info.
  5
      Future Response ======>13
Future Response =====>13
E 8
       Future Response ---->13
      Future Response ---->13
=
       Future Response =====>13
       Future Response =====>13
       Future Response =====>13
       Future Response ======>13
       Future Response ======>13
      Future Response =====>13
       Process finished with exit code 8
```

#### Consumer

```
Run: ArkaSyrcConsumer

| Austrians | Austr
```



# Exercise 20: Implementing a Custom Partitioning Strategy

We can create a class implementing the org.apache.kafka.clients.producer.Partitioner interface. This custom Partitioner will implement the business logic to decide where messages are sent.

**Use Case**: Let's jump to the next level by writing another program that implements customized message partitioning. The example consists of recollecting the IPs visiting a web site, which are recorded and published. The message has three parts: web site name, and IP address.

Let us create a topic with two partitions.

```
[cloudera@quickstart kafka]$ bin/kafka-topics.sh -create -zookeeper localhust:2181 -replication factor 1 -partitions 2 -topic ipHits SLF4]: Class path contains multiple SLF4] bindings.
SLF4]: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4]12-1,7.21.jar:/org/slf4]/impl/StaticLoggerBinder.class]
SLF4]: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4]12-1,7.5.jar:/org/slf4]/impl/StaticLoggerBinder.class]
SLF4]: See http://www.slf4].org/codes.htmlwmultiple bindings for an explanation.
SLF4]: Ketual binding is of type [org.slf4].impl.Log4]LoggerFactory]
Created topic 'ipHits'.
[cloudera@quickstart kafka]$ bin/kafka-topics.sh -describe -rookeeper localhost:2181 -topic ipHits
SLF4]: Class path contains multiple SLF4] bindings.
SLF4]: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4]12-1,7.21.jar:/org/slf4]/impl/StaticLoggerBinder.class]
SLF4]: Found binding in [jar:file:/usr/lib/kafka/libs/slf4]-log4]12-1,7.5.jar:/org/slf4]/impl/StaticLoggerBinder.class]
SLF4]: See http://www.slf4].org/codes.htmlwmultiple bindings for an explanation.
SLF4]: Actual binding is of type [org.slf4].impl.Log4]Log4]12-1,7.5.jar:/org/slf4]/impl/StaticLoggerBinder.class]
SLF4]: Actual binding is of type [org.slf4].impl.Log4]Log4]CoggerFactory]
Topic:ipHits Partition: 0 Leader: 0 Replicas: 0 Isr: 0
[cloudera@quickstart kafka]$
```

Let us create a CountryPartitioner that implements the org.apache.kafka.clients.producer.Partitioner interface as Scala class.

- 1. Create a SBT project and define all the dependencies.
- 2. Create a Scala class as CountryPartitioner

```
import kafka.producer._

import Scala.util

import org.apache.kafka.clients.producer.KafkaProducer

import org.apache.kafka.clients.producer.Partitioner

import org.apache.kafka.common.Cluster
```

```
object CountryPartitioner {
private var producer: KafkaProducer[String, String] = _
}
class CountryPartitioner extends Partitioner {
 def partition(key: AnyRef, a_numPartitions: Int): Int = {
  var partition = 1
  val partitionKey = key.asInstanceOf[String]
  val offset = partitionKey
  if (offset == "USA") {
   partition = 0
   println("its"+offset +"and its partition is "+partition)
  } else if (offset == "INDIA") {
   partition = 1
   println("its"+offset +"and its partition is "+partition)
  }
  partition
}
 override def partition(topic: String,
              key: AnyRef,
              keyBytes: Array[Byte],
              value: AnyRef,
              valueBytes: Array[Byte],
              cluster: Cluster): Int = partition(key, 10)
```

```
override def close() {
}

override def configure(configs: util.Map[String, _]) {
}
```

**Note**: We set a config property with a key equal to the value of **ProducerConfig.PARTITIONER\_CLASS\_CONFIG**, which matches the fully qualified name of our **CountryPartitioner** class. We also set countryName to partitionId, thus mapping the properties that we want to pass to CountryPartitioner.

3. Now let us create the Producer class

```
import Scala.util.Properties
import org.apache.kafka.clients.producer._

object KafkaProducerwithPartitioner extends App {
  val props = new Properties()
  props.put("bootstrap.servers", "localhost:9092")
  props.put("acks","1")
  props.put("key.serializer", "org.apache.kafka.common.serialization.StringSerializer")
  props.put("value.serializer", "org.apache.kafka.common.serialization.StringSerializer")

props.put("partitioner.class","CountryPartitioner")

val producer = new KafkaProducer[String, String](props)
  println("Start-- sending messages")
  var ipCountry="INDIA"
```

```
for(i <- 1 to 100) {
  if( i \%2 == 0){
   ipCountry = "USA"
  }
  else{
   ipCountry = "INDIA"
  }
  val record = new ProducerRecord[String, String]("ipHits", ipCountry,"Message" +
ipCountry)
  producer.send(record)
  println("Sent ->Message" + i)
}
 println("Sent--closing connection")
 producer.close()
}
```

4. Let us create one partitioned consumer which will consume all messages from partition 1.

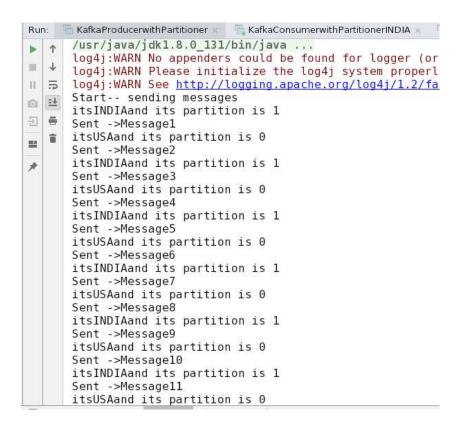
```
import org.apache.kafka.clients.consumer.ConsumerRecords
import org.apache.kafka.clients.consumer.KafkaConsumer
```

```
import Scala.util.{Collections, Properties}
import org.apache.kafka.common.TopicPartition
import scala.collection.ScalaConverters._
object KafkaConsumerwithPartitionerINDIA {
 def main(args: Array[String]): Unit = {
  val properties = new Properties()
  properties.put("bootstrap.servers", "localhost:9092")
  properties.put("group.id", "KafkaExampleNewConsumer2")
  properties.put("key.deserializer",
"org.apache.kafka.common.serialization.StringDeserializer")
  properties.put("value.deserializer",
"org.apache.kafka.common.serialization.StringDeserializer")
  properties.put("auto.offset.reset", "latest")
  val consumer = new KafkaConsumer[String, String](properties)
  //consumer.subscribe(util.Arrays.asList("testKafka"))
  val topicPartition = new TopicPartition("ipHits",1)
  consumer.assign(Collections.singletonList(topicPartition))
  while (true) {
   val records: ConsumerRecords[String, String] = consumer.poll(1000)
   records.asScala.foreach(record => println(s"Received message Partition:
"+record.partition() +"Offset:" +record.offset() +"Value "+record.value()))
  }
 }
```

5. Let us create one partitioned consumer which will consume all messages from partition 0.

```
import org.apache.kafka.clients.consumer._
import Scala.util.{Collections, Properties}
import org.apache.kafka.common.TopicPartition
import scala.collection.ScalaConverters.
object KafkaConsumerwithPartitionerUSA {
 def main(args: Array[String]): Unit = {
 val properties = new Properties()
  properties.put("bootstrap.servers", "localhost:9092")
  properties.put("group.id", "KafkaExampleNewConsumer")
  properties.put("key.deserializer",
"org.apache.kafka.common.serialization.StringDeserializer")
  properties.put("value.deserializer",
"org.apache.kafka.common.serialization.StringDeserializer")
  properties.put("auto.offset.reset", "latest")
  val consumer = new KafkaConsumer[String, String](properties)
  val topicPartition = new TopicPartition("ipHits",0)
  consumer.assign(Collections.singletonList(topicPartition))
   while (true) {
    val records: ConsumerRecords[String, String] = consumer.poll(1000)
    records.asScala.foreach(record => println(s"Received message Partition:
"+record.partition() +"Offset:" +record.offset() +"Value "+record.value()))
   }
  }
```

6. Start a producer as Scala Application.



Note: Observe the value of the offset for message send to each partition.

Start the consumers as Scala Application.
 Navigate to <consumer-console>. Consumer for partition 0.

```
Kakaroducewethrattooner Kakaconsumenethrattoonerstoone
/usr/jewa/jdkl.8.0_131/bin/jewa ...
log4j:WARN No appenders could be found for logger (org.apache.kafka.clients.consumer.Cons
log4j:WARN Please initialize the log4j system properly.
log4j:WARN See http://logging.apache.org/log4j/1.2/faq.html#nuconfig for more info.
Received message Partition: 80ffset:56Value MessageUSA
Received message Partition: 80ffset:51Value MessageUSA
Received message Partition: 00ffset:52Value MessageUSA
Received message Partition: 00ffset:54Value MessageUSA
Received message Partition: 80ffset:55Value MessageUSA
Received message Partition: 80ffset:55Value MessageUSA
Received message Partition: 80ffset:55Value MessageUSA
                                                                                       KafkaConsumerwithPartitioner#IDIA RafkaConsum
four:

    KafkaProducerwithPartitioner

 4 1
 # +
 11 5
 n H
        -
 1
         *
 12
                   Received message Partition:
Received message Partition:
                                                                                                 00ffset:55Value MessageUSA
00ffset:56Value MessageUSA
  p
                                                                                                 00ffset:57Value MessageUSA
00ffset:58Value MessageUSA
00ffset:59Value MessageUSA
                   Received message Partition:
                   Received message Partition:
Received message Partition:
                   Received message Partition:
                                                                                                 00ffset:60Value MessageUSA
00ffset:61Value MessageUSA
                   Received message Partition:
Received message Partition:
                                                                                                  00ffset:62Value MessageUSA
                   Received message Partition:
Received message Partition:
                                                                                                 00ffset:63Value MessageUSA
00ffset:64Value MessageUSA
                                                                                                 00ffset:65Value MessageUSA
00ffset:66Value MessageUSA
00ffset:67Value MessageUSA
                   Received message Partition:
                   Received message Partition:
Received message Partition:
                   Received message Partition: OUTset:68Value MessageUSA
Received message Partition: OUTset:69Value MessageUSA
Received message Partition: OUTset:70Value MessageUSA
Received message Partition: OUTset:70Value MessageUSA
Received message Partition: OUTset:72Value MessageUSA
Received message Partition: OUTset:72Value MessageUSA
                   Received message Partition: 00ffset:73Value MessageUSA
```

#### Consumer for partition 1.

```
| EnhancementPartition | Reflection | Reflec
```



## **Exercise 17: Spark Streaming**

Let's start with simple streaming example, to count the number of words in text data received from a data server listening on a TCP socket.

Create a DStream that represents streaming data from a TCP source, specified as hostname (e.g. localhost) and port (e.g. 9999).

Create a Scala project with below inputs:

```
import org.apache.spark.SparkConf
import org.apache.spark.SparkContext.
import org.apache.spark.streaming.StreamingContext
import org.apache.spark.streaming.Seconds
import org.apache.spark.storage.StorageLevel;
import org.apache.spark.streaming.StreamingContext.
object SparkStreamReceiver {
  def main(args: Array[String]): Unit = {
    var port:Int = 9999;
    val conf = new SparkConf()
                .setMaster("local[*]")
                .setAppName("Simple spark streaming")
    val ssc = new StreamingContext(conf, Seconds(3));
    val dstream = ssc.socketTextStream("localhost", port,
StorageLevel.MEMORY ONLY);
    val words = dstream.flatMap( .split(" "))
    val pairs = words.map(word => (word, 1))
    val wordcounts = pairs.reduceByKey(_ + _)
    wordcounts.print();
    ssc.start();
    ssc.awaitTermination();
```

\_\_\_\_\_

Export the jar file i.e. StreamApp.jar

```
[cloudera@quickstart workspace]$ ls -ltr
total 28
drwxrwxr-x 5 cloudera cloudera 4096 Sep 13 06:23 Wordcount
-rw-rw-r-- 1 cloudera cloudera 6055 Sep 13 06:40 wordcount.jar
drwxrwxr-x 7 cloudera cloudera 4096 Sep 13 07:33 StreamApp
drwxrwxr-x 8 cloudera cloudera 4096 Sep 13 07:33 WordCount
-rw-rw-r-- 1 cloudera cloudera 6920 Sep 13 07:45 StreamApp.jar
[cloudera@quickstart workspace]$ ■
```

Run the program using spark-submit

Run Netcat command and open another terminal to submit the job

```
nc -nlk 9999
```

```
[cloudera@quickstart workspace]$ nc -nlk 9999
hello world
hello hadoop
```

```
bin/spark-submit -master yarn --class solution.StreamingExample /home/cloudera/eclipse-workspace/Lesson4/target/sfpdapp-1.0.jar
```

Any lines typed in the terminal running the netcat server will be counted and printed on screen every second. It will look something like the following

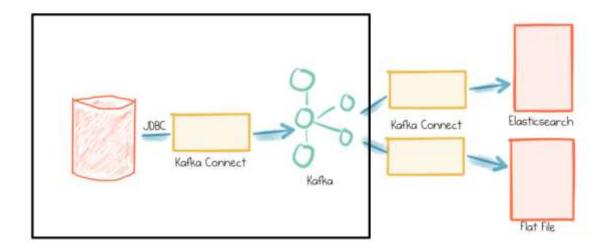
16/09/13 08:19:03 INFO executor.Executor: Finished task 0.0 in stage 4.0 (TID 3). 1312 bytes result sent to driver
16/09/13 08:19:03 INFO scheduler.TaskSetManager: Finished task 0.0 in stage 4.0 (TID 3) in 10 ms on localhost (1/1)
16/09/13 08:19:03 INFO scheduler.TaskSchedulerImpl: Removed TaskSet 4.0, whose t asks have all completed, from pool
16/09/13 08:19:03 INFO scheduler.DAGScheduler: ResultStage 4 (print at SparkStre amReceiver.scala:24) finished in 0.012 s
16/09/13 08:19:03 INFO scheduler.DAGScheduler: Job 2 finished: print at SparkStreamReceiver.scala:24, took 0.033085 s

Time: 1473779943000 ms

(hello,2)
(world,1)
(hadoop,1)



# Exercise 18: Apache Kafka Connector Example – Import Data from MySQL into Kafka



We are going to take care of the black box, as surrounded in the diagram above.

The Confluent JDBC Connector for Kafka Connect enables you to stream data to and from Kafka and any RDBMS that supports JDBC (which is to say pretty much any). It can stream entire schemas or just individual tables.

Pre-requisites to work with Kafka connect

- 1. Confluent should be installed
- 2. Java version 1.8 should be installed
- 3. Mysql database should be installed
- 4. Kafka and Schema Registry are running locally on the default ports.

Steps to study the functionality of jdbc-connector using kafka connect

1. cd/home/cloudera/training/confluent-4.1.0

bin/confluent start schema-registry

2. Connect to MySQL with user root and password cloudera

[cloudera@quickstart ~]\$ mysql -u root -p

```
[cloudera@quickstart ~]$ sudo mysql -u root -p
Enter password:
Welcome to the MySQL monitor. Commands end with ; or \g.
Your MySQL connection id is 15
Server version: 5.1.73 Source distribution

Copyright (c) 2000, 2013, Oracle and/or its affiliates. All rights reserved.

Oracle is a registered trademark of Oracle Corporation and/or its affiliates. Other names may be trademarks of their respective owners.

Type 'help;' or '\h' for help. Type '\c' to clear the current input statement.

mysql> ■
```

3. Next create a user, database, and a table:

mysql> create database kafkaconnect; Query OK, 1 row affected (0.00 sec) mysql> use kafkaconnect;

Database changed

mysql> CREATE TABLE authors(id INTEGER PRIMARY KEY

AUTO\_INCREMENT NOT NULL, name VARCHAR(255));

mysql> INSERT INTO authors(name) VALUES('Manish');

mysql> INSERT INTO authors(name) VALUES('Chanchal');

**Note**: If the table exists in the database, drop the table.

Note: Please note that this will not work on MySQL versions earlier than 5.6

- 4. Loading the jdbc connector
  - a) Checking the predefined connectors in confluent using the below command:
  - b) Navigate to the location where confluent is installed.

#### bin/confluent list connectors

```
[cloudera@quickstart confluent-4.1.0]$ bin/confluent list connectors
Bundled Predefined Connectors (edit configuration under etc/):
    elasticsearch-sink
    file-source
    file-sink
    jdbc-source
    jdbc-sink
    hdfs-sink
    s3-sink
[cloudera@quickstart confluent-4.1.0]$
```

c) Load the schema registry with the command to start schema registry on port 8081:

from the location /home/cloudera/training/confluent-oss-4.0.0-2.11.tar/confluent-4.0.0 (navigate to the location)

bin/connect-standalone/home/cloudera/Desktop/SW/confluent-5.5.0/etc/schema-registry/connect-avro-standalone.properties/home/cloudera/Desktop/SW/confluent-5.5.0/etc/schema-registry/quickstart-mysql.properties

5. Download the mysql connector java jar from below location:

Download the jar from https://mvnrepository.com/artifact/mysql/mysqlconnector-java/5.1.23

Click on the highlighted link "jar" which enables downloading the jar(yellow)



## mysql/mysql-connector-java/5.1.23

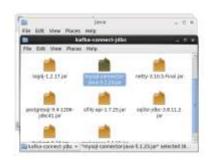
COPYING	2015-05-08 10:55	18122
COPYING.md5	2015-05-08 10:55	42
COPYING.shal	2015-05-08 10:55	50
mysql-connector-java-5.1.23-sources.jar	2015-05-08 10:55	767153
mysql-connector-java-5.1.23-sources.jar.md5	2015-05-08 10:55	74
mysql-connector-java-5.1.23-sources.jar.shal	2015-05-08 10:55	82
mysql-connector-java-5.1.23.jar	2015-05-08 10:55	843090
mysql-connector-java-5.1.23.jar.md5	2015-05-08 10:55	66
mysql-connector-java-5.1.23.jar.shal	2015-05-08 10:55	74
mysql-connector-java-5.1.23.pom	2015-05-08 10:55	1105
mysql-connector-java-5.1.23.pom.md5	2015-05-08 10:55	66
mysql-connector-java-5.1.23.pom.shal	2015-05-08 10:55	74

Note: As part of the lab, the jar is already downloaded and is kept in the location.

6. Copy the above MySQL Connector Jar in the following location:

/home/cloudera/training/confluent-4.1.0/share/java/kafka-connect-jdbc





7. Create a quickstart-mysql.properties in the location </home/cloudera/training/confluent-4.1.0/share/java/kafka-connect-jdbc> with the below content:

name=test-mysql-jdbc-autoincrement

connector. class = io. confluent. connect. jdbc. Jdbc Source Connector

tasks.max=1

connection.url=jdbc:mysql://localhost:3306/kafkaconnect?user=root&password=cloudera

mode=incrementing

incrementing.column.name=id

topic.prefix=test-

## Notes:

connection.url: In the connector configuration there are no security parameters. This is because SSL is not part of the JDBC standard and will depend on the JDBC driver in use. In general, you will need to configure SSL via the connection.url parameter which this jdbc driver connection url for mysql.

table.whitelist: is the configuration for setting the table which we want to copy in kafka

if autoincrementing column is set in a table, the column has to be specified.

topic-prefix: Prefix to prepend to table names to generate the name of the Kafka topic to publish data to.

| Industrial production of control of the control o

8. To check the data in kafka, run the Connect Avro console consumer

 $bin/kafka-avro\text{-}console\text{-}consumer\text{--}bootstrap\text{-}server\text{-}localhost\text{:}9092\text{--}topic\text{-}test-authors\text{--}from\text{-}beginning}$ 

"C(base) [clouderagquickstart confluent-5.5.0]\$ bin/kafka-avro-console-consumer --bootstrap-server localboot:9092 --topic test-authors --from-beginning {"id":1, "name":{"string":"Chanchat":}
{'id":2, "name":{"string":"Chanchat":}
\*\*CProcessed a total of 2 messages
{base} [cloudera@quickstart confluent-5.5.0]\$ ||

9. Add another record via the mysql command prompt:

## mysql INSERT INTO authors(name) VALUES('Rohan');

You can switch back to the console consumer and see the new record is added and, importantly, the old entries are not repeated:

Note: that the default polling interval is five seconds, so it may take a few seconds to show up. Depending on your expected rate of updates or desired latency, a smaller poll interval could be used to deliver updates more quickly.



# <u>Exercise 19: Apache Kafka Connector Example – Kafka Connect JDBC Sink</u>

#### Prerequisites:

- Confluent should be installed
- Java version 1.8 should be installed
- Mysql database should be installed
- Kafka and Schema Registry are running locally on the default ports.
- MySQL connector is installed.
- To Configure Data Sink Properties- Add a file as sink-quickstartmysql.properties in the location etc/kafka-connect-jdbc.
   The content is

```
name=test-sink-mysql-jdbc-autoincrement
connector.class=io.confluent.connect.jdbc.JdbcSinkConnector
tasks.max=1
catalog=training
topics=orders
connection.url=jdbc:mysql://localhost:3306/training?user=root&password=clouder
a
auto.create=true
```

2. Let us Start standalone connector, with the command in terminal

 $bin/connect\-standalone\ etc/schema\-registry/connect\-avro\-standalone.properties\ /home/cloudera/Desktop/SW/confluent\-5.5.0/etc/schema\-registry/sink\-quick start-mysql.properties$ 

#### Observe the logs, the table is created

3. Let us Produce some record into the orders topic, passing the schema details of the table which will be automatically created

```
bin/kafka-avro-console-producer --broker-list localhost:9092 --topic orders --
property
value.schema='{"type":"record","name":"myrecord","fields":[{"name":"id","type":"in
```

```
t"},{"name":"product", "type": "string"}, {"name":"quantity", "type": "int"}, {"name":"price","type": "float"}]}'
```

The console producer is waiting for input. Copy and paste the following record into the terminal and press **ENTER** 

```
{"id": 999, "product": "foo", "quantity": 100, "price": 50}

{"id": 999, "product": "foo", "quantity": 100, "price": 50}

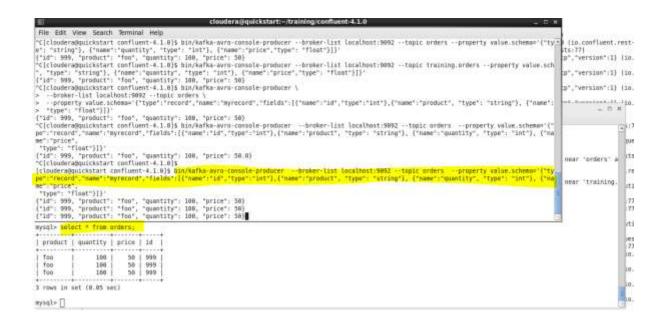
{"id": 999, "product": "foo", "quantity": 100, "price": 50}
```

4. Now if we query the database, we will see that the orders table was automatically created and contains the record.

```
(base) [cloudera@quickstart spark]$ hdfs dfs -ls /user/hive/warehouse
Found 7 items
drwxrwxrwx
            - cloudera supergroup
                                             0 2020-06-03 06:03 /user/hive/warehouse/bdp.db
drwxrwxrwx
            - cloudera supergroup
                                             0 2020-06-04 01:03 /user/hive/warehouse/demohivespark.db
            - cloudera supergroup
                                             0 2020-06-05 23:51 /user/hive/warehouse/hive_part_tbl
drwxrwxrwx
drwxr-xr-x

    cloudera supergroup

                                             0 2020-06-05 23:50 /user/hive/warehouse/hive records
drwxrwxrwx
            - cloudera supergroup
                                             0 2020-06-05 04:55 /user/hive/warehouse/partitioned_user
            - cloudera supergroup
- cloudera supergroup
drwxrwxrwx
                                             0 2020-06-05 23:49 /user/hive/warehouse/src
drwxrwxrwx
                                            0 2020-06-05 04:48 /user/hive/warehouse/users part
(base) [cloudera@quickstart spark]$
```





## Exercise 20: Add Scala Kernal in Jupyter

Open a terminal and navigate to root folder

cd ~

Download almond and scala libs

\$ curl -Lo coursier https://git.io/coursier-cli && chmod +x coursier

- \$ ./coursier bootstrap \
  - -r jitpack \
  - i user ·I user:sh.almond:scala-kernel-api\_2.12.8:0.7.0 \
  - sh.almond:scala-kernel 2.12.8:0.7.0 \

#### -o almond

> Add Scala kernel to Jupyter

```
$ ./almond --install --id scala_2_12_8 --display-name "Scala 2.12.8"
```

➤ List kernelsPermalink

```
jupyter kernelspec list
```

```
(base) [cloudera@quickstart ~]$ jupyter kernelspec list

Available kernels:

scala_2_12_8 /home/cloudera/.local/share/jupyter/kernels/scala_2_12_8

python2 /home/cloudera/anaconda2/share/jupyter/kernels/python2

(base) [cloudera@quickstart ~]$ ■
```

Launch pysparknb to launch jupyter notebook from the terminal.

```
(base) [cloudera@quickstart spark]$ pysparknb
[I 01:18:05.336 NotebookApp] The port 8888 is already in use, trying another port.
[I 01:18:05.357 NotebookApp] Loading IPython parallel extension
[I 01:18:05.397 NotebookApp] JupyterLab extension loaded from /home/cloudera/anaconda2/lib/python2.7/site-packages/jupyterlab
[I 01:18:05.397 NotebookApp] JupyterLab application directory is /home/cloudera/anaconda2/share/jupyter/lab
[I 01:18:05.408 NotebookApp] Serving notebooks from local directory; /home/cloudera/Desktop/SW/spark
[I 01:18:05.408 NotebookApp] The Jupyter Notebook is running at:
[I 01:18:05.408 NotebookApp] http://localhost:8889/
[I 01:18:05.408 NotebookApp] Use Control-C to stop this server and shut down all kernels (twice to skip confirmation).
```

Its running, open mozilla with the provided url.

To run Zeppelin and create a Zeppelin notebook, run the following command: zeppelin-daemon.sh start

```
File Edit View Search Terminal Help

(base) [cloudera@quickstart zeppelin-0.8.2-bin-all]$ bin/zeppelin-daemon.sh start

Log dir doesn't exist, create /home/cloudera/Desktop/SW/zeppelin-0.8.2-bin-all/logs

Pid dir doesn't exist, create /home/cloudera/Desktop/SW/zeppelin-0.8.2-bin-all/run

Zeppelin start

(base) [cloudera@quickstart zeppelin-0.8.2-bin-all]$
```

There are some useful Zeppelin commands one should know:

- $\$  zeppelin-daemon.sh start -> To start the Daemon
- $\$  zeppelin-daemon.sh stop -> To stop the Daemon
- $\$  zeppelin-daemon.sh restart -> To restart the Daemon

Your Zeppelin Notebook should be accessible from the following link :  $\label{eq:htp:/localhost:8080/} http://localhost:8080/$