

HDP Developer: Enterprise Apache Spark 1 Scala Lab Guide Guide

Rev 1





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Lab 0: Pre-lab Setup

About This Lab

Objective:

Set up the lab environment and confirm functionality

File Locations:

N/A

Successful Outcome:

User will set up the HDP cluster and verify login

Before You Begin:

Connect to the lab Environment

Lab Steps

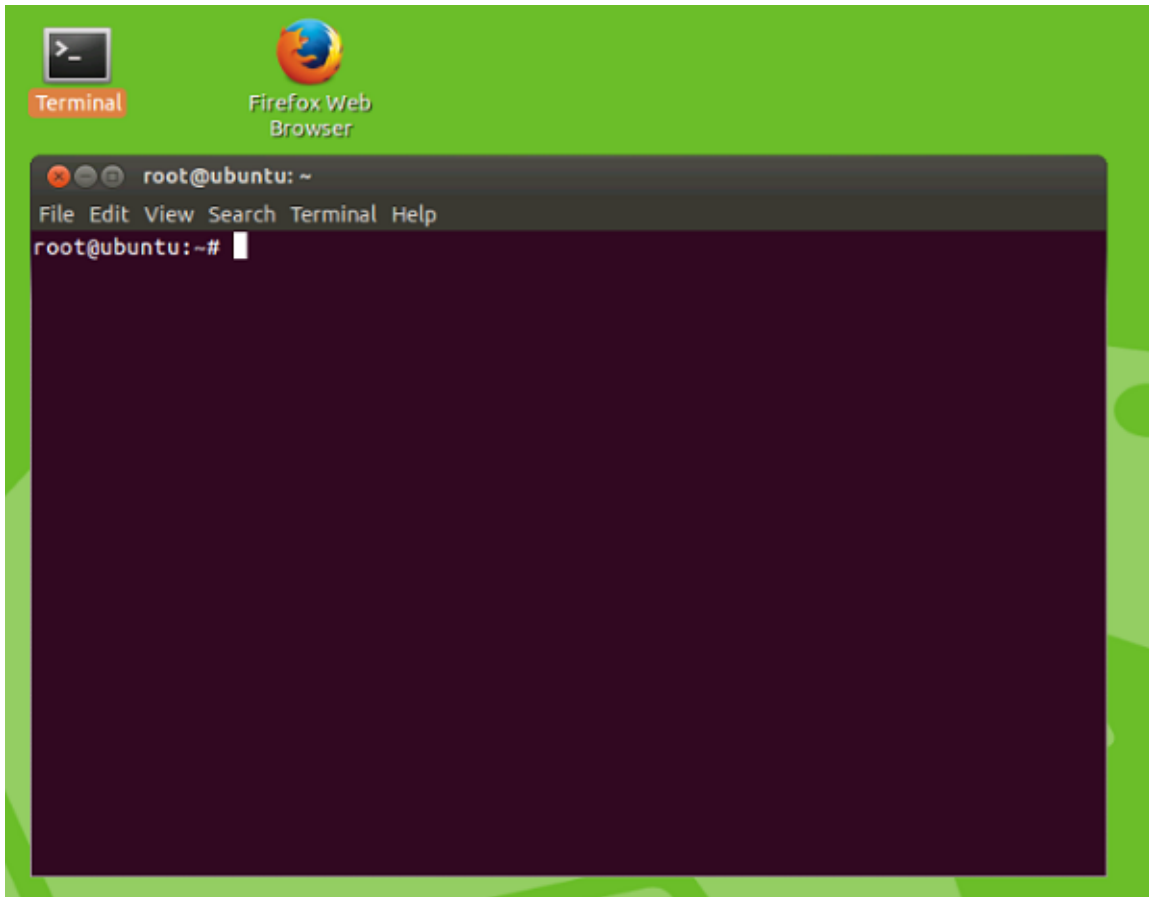
Perform the following steps:

1. **Start the HDP cluster.**
 - a. Connect to the lab environment.



- b. Double-click on the Terminal icon on the desktop.

Lab 0: Pre-lab Setup



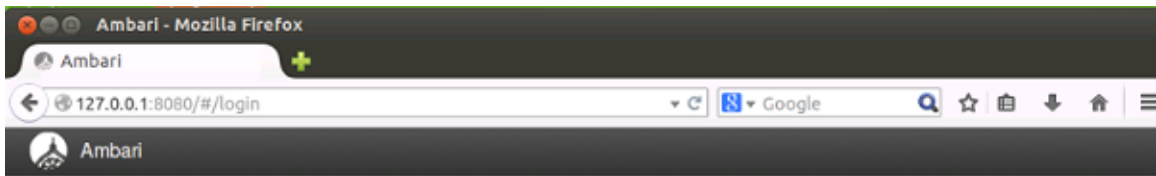
- c. Use SSH to connect to the Docker container – named “sandbox” – that has been a single-node HDP cluster installation configured.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
Warning: Permanently added the RSA host key for IP address '172.17.0.1' to the
list of known hosts.
Last login: Thu Apr 21 23:53:04 2016 from ip-172-17-0-1.ec2.internal
```

2. Verify and, if necessary, start HDP cluster services.

- a. Open a Firefox web browser and log into the Ambari Web UI using <http://sandbox:8080>.



Sign in

Username

Password

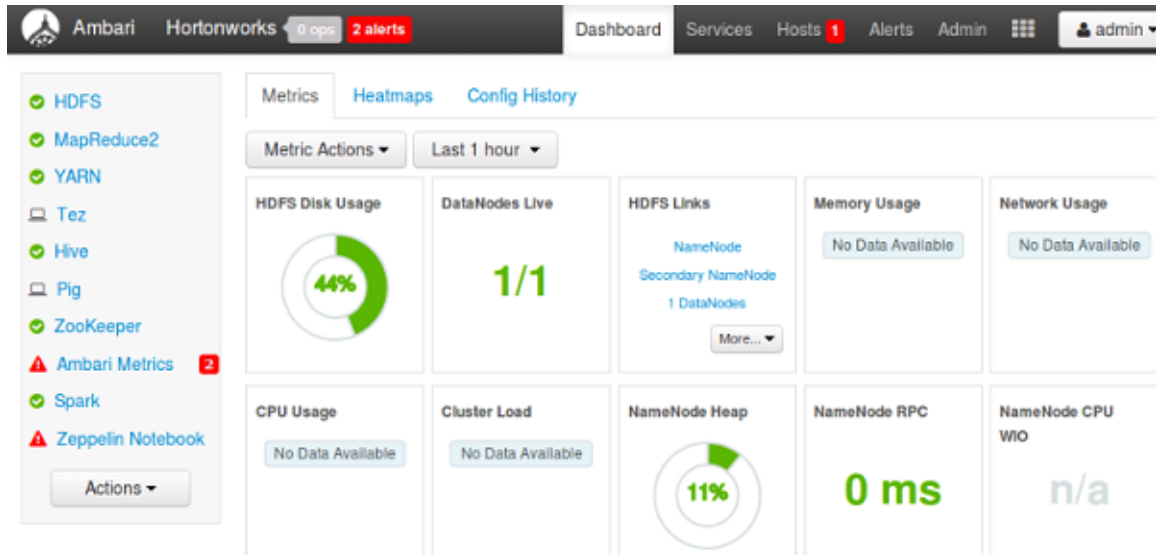
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- b. Supply a username and password of admin and admin, then click the Sign in button to get to the Ambari Web UI dashboard.

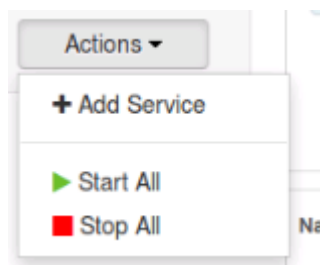
Username

Password

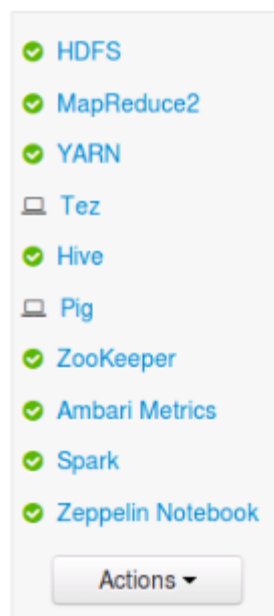
Lab 0: Pre-lab Setup



- c. All services should be running. If not, start any stopped services by clicking on the Actions button at the bottom left and selecting Start All.



- d. If a restart was necessary, give the services a couple of minutes to start. One or more of them may initially report failure, but after waiting will go green. When everything has settled, your dashboard list of services should look similar to this:



3. Confirm HDFS (Hadoop Distributed File System) access from the command line.

- a. Go back to the terminal window that is connected to the sandbox Docker container (reopen and reconnect if necessary) and switch users so that you can run HDFS administrative commands.

```
# su hdfs
```

```
[root@sandbox ~]# su hdfs  
[hdfs@sandbox root]#
```

- b. To verify HDFS connectivity, run the `hdfs dfsadmin -report` command. Verify that it provides output similar to the screenshot provided.

```
# hdfs dfsadmin -report
```

```
[hdfs@sandbox root]# hdfs dfsadmin -report  
Configured Capacity: 100000174080 (93.13 GB)  
Present Capacity: 57822167040 (53.85 GB)  
DFS Remaining: 56309686272 (52.44 GB)  
DFS Used: 1512480768 (1.41 GB)  
DFS Used%: 2.62%  
Under replicated blocks: 82  
Blocks with corrupt replicas: 0  
Missing blocks: 0  
Missing blocks (with replication factor 1): 0  
  
-----  
Live datanodes (1):  
  
Name: 172.17.0.1:50010 (sandbox)  
Hostname: sandbox  
Decommission Status : Normal  
Configured Capacity: 100000174080 (93.13 GB)  
DFS Used: 1512480768 (1.41 GB)  
Non DFS Used: 42178007040 (39.28 GB)  
DFS Remaining: 56309686272 (52.44 GB)
```

- c. Exit the HDFS administrative user and go back to being the root user.


```
# exit
```

```
[hdfs@sandbox root]# exit  
exit  
[root@sandbox ~]#
```

- d. Run the `jps` command and verify that a process called NameNode is running.

```
# jps
```

```
[root@sandbox ~]# jps
3199 JobHistoryServer
19233 -- process information unavailable
2577 SecondaryNameNode
4541 HistoryServer
4100 RunJar
3090 ApplicationHistoryServer
4415 RunJar
3303 NodeManager
2389 QuorumPeerMain
18604 -- process information unavailable
12122 ZeppelinServer
18406 Jps
2284 AmbariServer
2697 DataNode
11056 HMaster
2461 NameNode
11097 ApplicationHistoryServer
3860 RunJar
2995 ResourceManager
[root@sandbox ~]#
```



Result

You have successfully connected to your lab environment, used SSH to connect to the HDP cluster Docker container, started Ambari and all HDP services, and verified connection to HDFS and operation of the NameNode process.

Lab 1: Using HDFS Commands

About This Lab

Objective:

View, add, manipulate, and remove files and directories to and from HDFS using `hdfs dfs` commands.

File Locations:

`/root/spark/data/`

Successful Outcome:

You will have added, manipulated, and deleted several files and folders in HDFS

Before You Begin:

You should be logged in to your lab environment

Lab Steps

Perform the following steps:

1. **View the `hdfs dfs` command.**

- a. Open a Terminal window and use `ssh` to connect to the sandbox virtual machine.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
Last login: Thu May 19 15:55:24 2016 from ip-172-17-42-1.ec2.internal
[root@sandbox ~]#
```

- b. From the command line, enter the `hdfs dfs` command with no arguments to view its usage.

```
# hdfs dfs
```

```

[root@sandbox ~]# hdfs dfs
Usage: hadoop fs [generic options]
  [-appendToFile <localsrc> ... <dst>]
  [-cat [-ignoreCrc] <src> ...]
  [-checksum <src> ...]
  [-chgrp [-R] GROUP PATH...]
  [-chmod [-R] <MODE[,MODE]... | OCTALMODE> PATH...]
  [-chown [-R] [OWNER][:[GROUP]] PATH...]
  [-copyFromLocal [-f] [-p] [-l] <localsrc> ... <dst>]
  [-copyToLocal [-p] [-ignoreCrc] [-crc] <src> ... <localdst>]
  [-count [-q] [-h] [-v] [-t [<storage type>]] <path> ...]
  [-cp [-f] [-p | -p[topax]] <src> ... <dst>]
  [-createSnapshot <snapshotDir> [<snapshotName>]]
  [-deleteSnapshot <snapshotDir> <snapshotName>]
  [-df [-h] [<path> ...]]
  [-du [-s] [-h] <path> ...]
  [-expunge]
  [-find <path> ... <expression> ...]
  [-get [-p] [-ignoreCrc] [-crc] <src> ... <localdst>]
  [-getfacl [-R] <path>]
  [-getfattr [-R] {-n name | -d} [-e en] <path>]
  [-getmerge [-nl] <src> <localdst>]
  [-help [cmd ...]]
  [-ls [-d] [-h] [-R] [<path> ...]]
  [-mkdir [-p] <path> ...]
  [-moveFromLocal <localsrc> ... <dst>]
  [-moveToLocal <src> <localdst>]
  [-mv <src> ... <dst>]
  [-put [-f] [-p] [-l] <localsrc> ... <dst>]
  [-renameSnapshot <snapshotDir> <oldName> <newName>]
  [-rm [-f] [-r|-R] [-skipTrash] [-safely] <src> ...]
  [-rmdir [--ignore-fail-on-non-empty] <dir> ...]
  [-setfacl [-R] [{-b|-k} {-m|-x <acl_spec>} <path>][[--set <acl_spec> <pa
th>]]
  [-setfattr {-n name [-v value] | -x name} <path>]
  [-setrep [-R] [-w] <rep> <path> ...]
  [-stat [format] <path> ...]
  [-tail [-f] <file>]
  [-test [-defsz] <path>]
  [-text [-ignoreCrc] <src> ...]
  [-touchz <path> ...]
  [-truncate [-w] <length> <path> ...]
  [-usage [cmd ...]]

Generic options supported are
  -conf <configuration file>      specify an application configuration file
  -D <property=value>             use value for given property
  -fs <local|namenode:port>       specify a namenode
  -jt <local|resourcemanager:port> specify a ResourceManager
  -files <comma separated list of files> specify comma separated files to be co
  pied to the map reduce cluster
  -libjars <comma separated list of jars> specify comma separated jar files to
  include in the classpath.
  -archives <comma separated list of archives> specify comma separated archives
  to be unarchived on the compute machines.

The general command line syntax is
bin/hadoop command [genericOptions] [commandOptions]

[root@sandbox ~]# █

```


2. Create directories in HDFS.

- a. Enter the `hdfs dfs -ls` command with no directory specified to view the contents of the current user's home directory in HDFS. Since you are logged in as the user `root`, the typical home directory location will be `/user/root`.

```
# hdfs dfs -ls
```

```
[root@sandbox ~]# hdfs dfs -ls
Found 10 items
drwx----- - root hdfs      0 2016-04-02 02:00 .Trash
drwxr-xr-x - root hdfs      0 2016-04-24 22:58 .hiveJars
drwxr-xr-x - root hdfs      0 2016-04-13 07:01 .sparkStaging
-rw-r--r-- 3 root hdfs    205888 2016-04-01 16:45 airports.csv
-rw-r--r-- 3 root hdfs    37794 2016-04-01 16:47 carriers.csv
drwxr-xr-x - root hdfs      0 2016-04-02 15:10 checkpointDir
-rw-r--r-- 3 root hdfs  136035258 2016-04-01 16:45 flights.csv
-rw-r--r-- 3 root hdfs    428796 2016-04-01 16:47 plane-data.csv
-rw-r--r-- 3 root hdfs     8596 2016-04-13 06:48 selfishgiants.txt
drwxr-xr-x - root hdfs      0 2016-04-01 15:01 test
[root@sandbox ~]#
```

- b. Run the command again, but this time specify the root folder for all of HDFS.

```
# hdfs dfs -ls /
```

```
[root@sandbox ~]# hdfs dfs -ls /
Found 9 items
drwxrwxrwx - yarn  hadoop      0 2016-04-22 01:53 /app-logs
drwxr-xr-x - hdfs  hdfs      0 2015-12-17 22:13 /apps
drwxr-xr-x - yarn  hadoop      0 2016-04-01 13:00 /ats
drwxr-xr-x - hdfs  hdfs      0 2015-12-02 10:30 /hdp
drwxr-xr-x - mapred hdfs      0 2015-12-02 10:30 /mapred
drwxrwxrwx - mapred hadoop      0 2015-12-02 10:30 /mr-history
drwxrwxrwx - spark  hadoop      0 2016-05-27 09:30 /spark-history
drwxrwxrwx - hdfs  hdfs      0 2016-04-25 12:12 /tmp
drwxr-xr-x - hdfs  hdfs      0 2015-12-17 22:13 /user
[root@sandbox ~]#
```

- c. Create a directory named `dirTest` in the current user's home directory in HDFS.

```
# hdfs dfs -mkdir dirTest
```

```
[root@sandbox ~]# hdfs dfs -mkdir dirTest
[root@sandbox ~]#
```

- d. Verify the folder was created successfully.

```
# hdfs dfs -mkdir dirTest
```

```
[root@sandbox ~]# hdfs dfs -ls
```

```
drwxr-xr-x  - root hdfs          0 2016-05-27 09:35 dirTest
```

- e. Verify that this directory was created in the user's home directory.

```
# hdfs dfs -ls /user/root
```

```
[root@sandbox ~]# hdfs dfs -ls /user/root
```

```
drwxr-xr-x  - root hdfs          0 2016-05-27 09:35 dirTest
```



NOTE:

There is no difference between performing the `-ls` command when you specify no directories and when you specify the user's home directory. All commands will be executed in the user's home directory unless otherwise specified.

- f. Use `-mkdir` to create subdirectory `dir1` in the `dirTest` directory. Then run the command again with the `-p` option to create an additional subdirectory, `dir2`, which also contains its own subdirectory, `dir3`.

```
# hdfs dfs -mkdir dirTest/dir1
```

```
# hdfs dfs -mkdir -p dirTest/dir2/dir3
```

```
[root@sandbox ~]# hdfs dfs -mkdir dirTest/dir1
[root@sandbox ~]# hdfs dfs -mkdir -p dirTest/dir2/dir3
[root@sandbox ~]#
```

- g. Run the `hdfs dfs -ls -R` command to recursively view the contents of the user's home directory, and verify that all three directories from the previous step were successfully created.

```
# hdfs dfs -ls -R
```

```
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest/dir1
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest/dir2
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest/dir2/dir3
```

3. Delete directories in HDFS.

- a. Delete the `dir1` directory and verify it no longer exists.

```
# hdfs dfs -rmdir dirTest/dir1
```

```
[root@sandbox ~]# hdfs dfs -rmdir dirTest/dir1
[root@sandbox ~]#
```

```
# hdfs dfs -ls dirTest
```

```
[root@sandbox ~]# hdfs dfs -ls dirTest
Found 1 items
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest/dir2
```

- b. This command works because the directory is empty. Run the command again, and this time try to delete the `dir2` directory and note the error message. Then verify that the directory still exists.

```
# hdfs dfs -rmdir dirTest/dir2
```

```
# hdfs dfs -ls dirTest
```

```
[root@sandbox ~]# hdfs dfs -rmdir dirTest/dir2
rmdir: 'dirTest/dir2': Directory is not empty
[root@sandbox ~]# hdfs dfs -ls dirTest
Found 1 items
drwxr-xr-x  - root hdfs          0 2016-05-27 12:48 dirTest/dir2
[root@sandbox ~]#
```

- c. To delete a directory and all of its contents, use `hdfs dfs -rm -R <directory path>`.



WARNING:

Be very careful not to run this without specifying a directory, as the default behavior would be to delete the user's home directory and all contents (in our case, the `/user/root` directory and everything it contains).

Use this command to delete the `dir2` directory and its contents, and verify that the directory has been deleted.

```
# hdfs dfs -rm -R dirTest/dir2
```

```
# hdfs dfs -ls dirTest
```

```
[root@sandbox ~]# hdfs dfs -rm -R dirTest/dir2
16/05/27 13:13:57 INFO fs.TrashPolicyDefault: Namenode trash configuration: Deletion interval = 360 minutes, Emptier interval = 0 minutes.
Moved: 'hdfs://sandbox:8020/user/root/dirTest/dir2' to trash at: hdfs://sandbox:8020/user/root/.Trash/Current
[root@sandbox ~]# hdfs dfs -ls dirTest
[root@sandbox ~]#
```

4. Upload, copy, and delete HDFS files.

- a. The sandbox container image should be preloaded with some test files. Change directories to `/root/spark/data/` and view the contents of this directory.

```
# cd /root/spark/data/
# ls
```

```
[root@sandbox ~]# cd /root/spark/data/
[root@sandbox data]# ls
airports.csv  data.txt      plane-data.csv  small_blocks.txt
carriers.csv  flights.csv   selfishgiant.txt spamEmail
```

- b. Put the `data.txt` file into the `dirTest` directory in HDFS.

```
# hdfs dfs -put data.txt dirTest/
```

```
[root@sandbox data]# hdfs dfs -put data.txt dirTest/
[root@sandbox data]#
```

- c. Verify the file was uploaded successfully.

```
# hdfs dfs -ls dirTest
```

```
[root@sandbox data]# hdfs dfs -ls dirTest
Found 1 items
-rw-r--r--   3 root hdfs      20 2016-05-27 13:22 dirTest/data.txt
[root@sandbox data]#
```

- d. Create a copy of the `data.txt` file named `datacopy.txt` and verify the operation was successful.

```
# hdfs dfs -cp dirTest/data.txt dirTest/datacopy.txt
# hdfs dfs -ls dirTest
```

Lab 1: Using HDFS Commands

```
[root@sandbox data]# hdfs dfs -cp dirTest/data.txt dirTest/datacopy.txt
[root@sandbox data]# hdfs dfs -ls dirTest
Found 2 items
-rw-r--r--   3 root hdfs      20 2016-05-27 13:22 dirTest/data.txt
-rw-r--r--   3 root hdfs      20 2016-05-27 13:28 dirTest/datacopy.txt
[root@sandbox data]#
```



QUESTION:

What do you think would have happened if the `dirTest` directory had not been explicitly specified as the location for the `datacopy.txt` file?

- e. Now delete the `datacopy.txt` file and verify it has been removed.

```
# hdfs dfs -rm dirTest/datacopy.txt
```

```
# hdfs dfs -ls dirTest
```

```
[root@sandbox data]# hdfs dfs -rm dirTest/datacopy.txt
16/05/27 13:32:41 INFO fs.TrashPolicyDefault: Namenode trash configuration: Deletion interval = 360 minutes, Emptier interval = 0 minutes.
Moved: 'hdfs://sandbox:8020/user/root/dirTest/datacopy.txt' to trash at: hdfs://sandbox:8020/user/root/.Trash/Current
[root@sandbox data]# hdfs dfs -ls dirTest
Found 1 items
-rw-r--r--   3 root hdfs      20 2016-05-27 13:22 dirTest/data.txt
[root@sandbox data]#
```

5. View, download, and download merged files in HDFS.

- a. View the contents of the `data.txt` file in HDFS.

```
# hdfs dfs -cat dirTest/data.txt
```

```
[root@sandbox data]# hdfs dfs -cat dirTest/data.txt
This is a test file
[root@sandbox data]#
```

OR

```
# hdfs dfs -tail dirTest/data.txt
```

```
[root@sandbox data]# hdfs dfs -tail dirTest/data.txt
This is a test file
[root@sandbox data]#
```

- b. Download the `data.txt` file from HDFS to the `/tmp` directory on the local file system and verify the operation was successful.

Lab 1: Using HDFS Commands

```
# hdfs dfs -get dirTest/data.txt /tmp
# ls /tmp/data*
```

```
[root@sandbox data]# hdfs dfs -get dirTest/data.txt /tmp
[root@sandbox data]# ls /tmp/data*
/tmp/data.txt
[root@sandbox data]#
```

- c. View the contents of the `small_blocks.txt` file on the local file system. It should be in the current directory.

```
# cat small_blocks.txt
```

```
[root@sandbox data]# cat small_blocks.txt
This is data in the small blocks file
[root@sandbox data]#
```

- d. Upload the `small_blocks.txt` into the `dirTest` folder in HDFS and verify that you now have two files in `dirTest`.

```
# hdfs dfs -put small_blocks.txt dirTest/
# hdfs dfs -ls dirTest
```

```
[root@sandbox data]# hdfs dfs -put small_blocks.txt dirTest/
[root@sandbox data]# hdfs dfs -ls dirTest
Found 2 items
-rw-r--r--   3 root hdfs      20 2016-05-27 13:22 dirTest/data.txt
-rw-r--r--   3 root hdfs      38 2016-05-27 13:48 dirTest/small_blocks.txt
[root@sandbox data]#
```

- e. Merge and download all of the contents of the `dirTest` directory in HDFS to a file named `merged.txt` in the `/tmp` directory on the local file system. Verify that the `merged.txt` file was successfully created.

```
# hdfs dfs -getmerge dirTest /tmp/merged.txt
# ls /tmp/merged*
```

```
[root@sandbox data]# hdfs dfs -getmerge dirTest /tmp/merged.txt
[root@sandbox data]# ls /tmp/merged*
/tmp/merged.txt
[root@sandbox data]#
```

View the contents of the `merged.txt` file to confirm that it contains the contents of both files that were in the `dirTest` directory.

```
# cat /tmp/merged.txt
```

Lab 1: Using HDFS Commands

```
[root@sandbox data]# cat /tmp/merged.txt
This is a test file
This is data in the small blocks file
[root@sandbox data]#
```

- f. Change directories back to the root user's home directory.

```
# cd ~
# pwd
```

```
[root@sandbox data]# cd ~
[root@sandbox ~]# pwd
/root
[root@sandbox ~]#
```

Result

You have successfully created, manipulated, and deleted files and directories in HDFS.

Lab 2: Introduction to Spark REPLs and Zeppelin

About This Lab

Objective:

Access and browse Spark REPLs and Zeppelin

File Locations:

N/A

Successful Outcome:

Use Spark REPLs and browse Zeppelin

Before You Begin:

Complete the Pre-Lab and confirm cluster operation

Lab Steps

Perform the following steps:

1. Access the Spark REPLs.

- a. Open a Terminal window and use ssh to connect to the sandbox virtual machine.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
Last login: Thu May 19 15:55:24 2016 from ip-172-17-42-1.ec2.internal
[root@sandbox ~]#
```

- b. Run the Spark REPL for Scala.

```
# spark-shell
```

```
[root@sandbox ~]# spark-shell
```

```
16/05/04 10:26:35 INFO metastore: Connected to metastore.
16/05/04 10:26:35 INFO SessionState: Created local directory: /tmp/0b672003-16b5-4a63-973f-ee6b35238448_resources
16/05/04 10:26:35 INFO SessionState: Created HDFS directory: /tmp/hive/root/0b672003-16b5-4a63-973f-ee6b35238448
16/05/04 10:26:35 INFO SessionState: Created local directory: /tmp/root/0b672003-16b5-4a63-973f-ee6b35238448
16/05/04 10:26:35 INFO SessionState: Created HDFS directory: /tmp/hive/root/0b672003-16b5-4a63-973f-ee6b35238448/_tmp_space.db
16/05/04 10:26:35 INFO SparkILoop: Created sql context (with Hive support)..
SQL context available as sqlContext.
scala>
```



```
>>> SC
<pyspark.context.SparkContext object at 0xd09390>
>>> sc.appName
u'PySparkShell'
>>> sc.version
u'1.6.0'
```

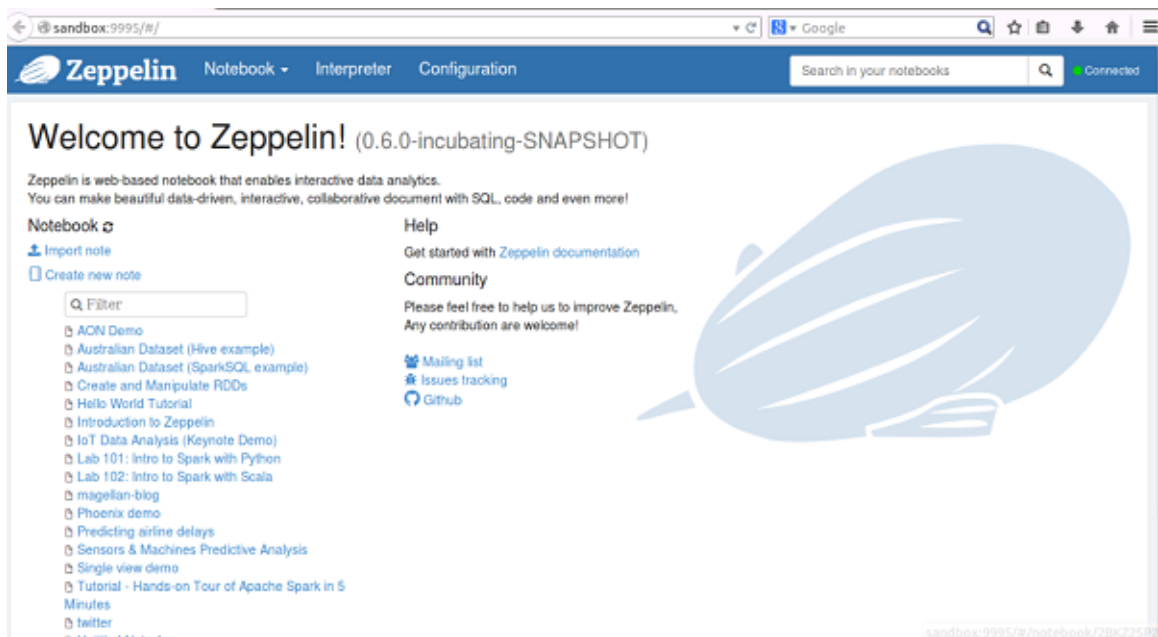
g. Exit the Spark Python REPL.

```
>>> exit()
```

```
>>> exit()
```

2. Access and browse Zeppelin.

a. Open the Firefox browser and enter the following URL to view the Zeppelin UI:
<http://sandbox:9995/>



- b. Click Interpreter in the top menu and note that Zeppelin's default interpreter is set to Spark and has a number of default settings configured.

The screenshot shows the Zeppelin web interface. At the top, there is a blue navigation bar with the Zeppelin logo and three menu items: 'Notebook', 'Interpreter', and 'Configuration'. Below the navigation bar, the page title is 'Interpreters'. Underneath, there is a subtitle: 'Manage interpreters settings. You can create create / remove settings. Note can bind/unbind these interpreter settings.' The main content area shows the configuration for the 'spark' interpreter, which is the default. It includes a section for 'Option' with a toggle for 'Separate Interpreter for each note' which is currently turned off. Below that is a 'Properties' section containing a table of configuration options.

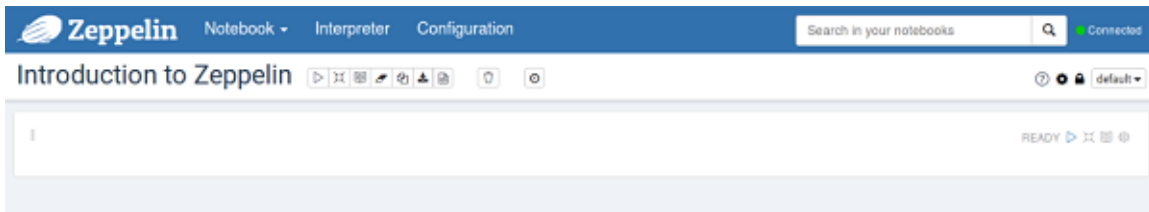
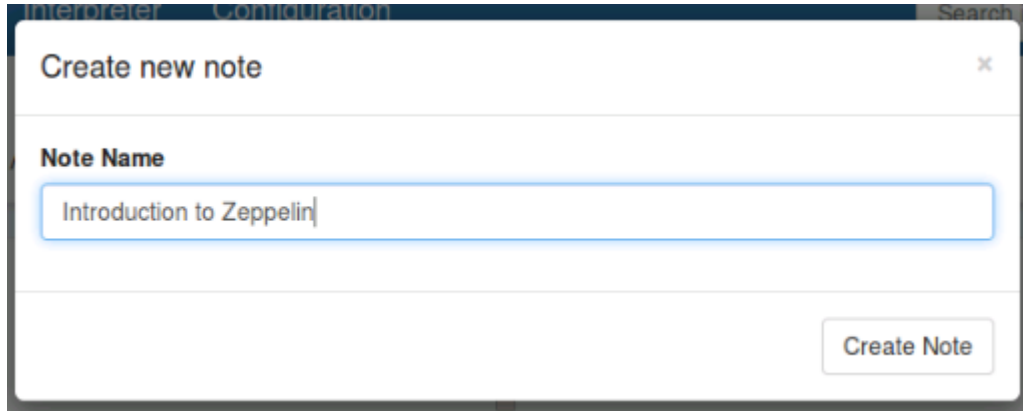
name	value
spark.cores.max	
zeppelin.spark.printREPLOutput	true
master	yarn-client
zeppelin.spark.maxResult	1000
zeppelin.dep.localrepo	local-repo
spark.app.name	Zeppelin
spark.executor.memory	512m

- c. Click on Notebook in the top menu and select Create new note from the resulting drop down options.

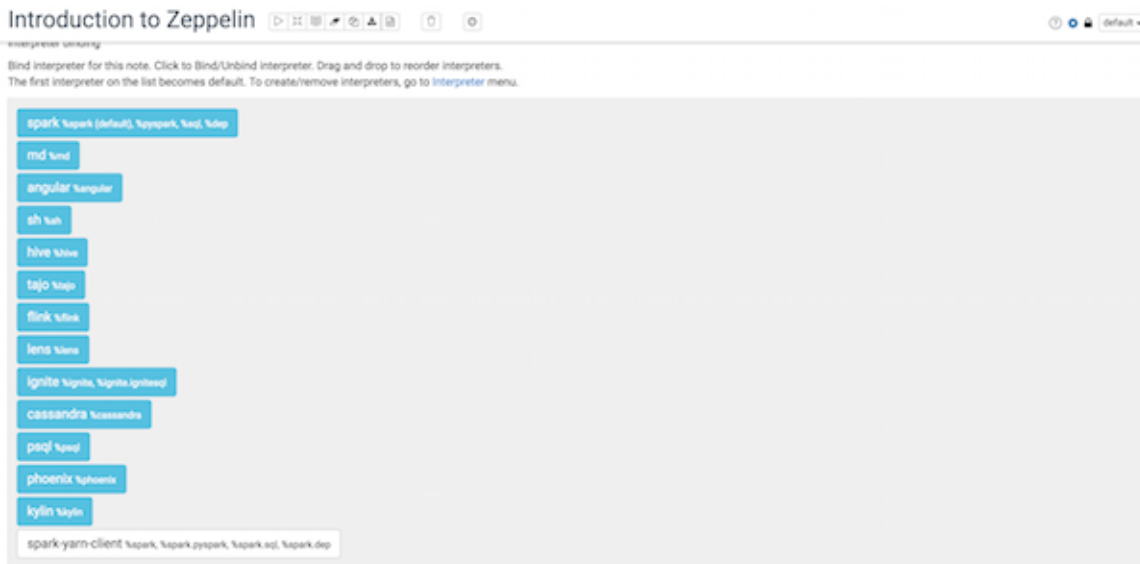
The screenshot shows the Zeppelin web interface with the 'Notebook' menu open. The top navigation bar is visible with the Zeppelin logo and the 'Notebook', 'Interpreter', and 'Configuration' menus. The 'Notebook' dropdown menu is expanded, showing a '+ Create new note' option at the top. Below this is a search box labeled 'Filter'. Underneath the search box, there are three options listed: 'AON Demo', 'Australian Dataset (Hive example)', and 'Australian Dataset (SparkSQL example)'. The background shows the 'Interpreters' page content, which is partially obscured by the dropdown menu.

Lab 2: Introduction to Spark REPLs and Zeppelin

- d. Name this note Introduction to Zeppelin and click Create Note.



- e. At the top right click on the gear icon to change interpreter binding. Your administrator has enabled an interpreter called “**spark yarn-client**” which is configured for the HDP cluster you are using. Drag it to the top of the list of interpreters, and click the Save button.



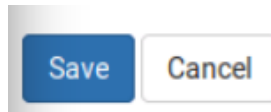
Lab 2: Introduction to Spark REPLs and Zeppelin

Introduction to Zeppelin ⏪ ⏩ ⏴ ⏵ ⏶ ⏷ ⏸ ⏹ ⏺ ⏻ ⏼ ⏽ ⏾ ⏿ 🔍 🌐 🏠 default ▾

interpreter: spark

Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu.

The first interpreter on the list is treated as the default interpreter. Scroll down to find the Save button.



- f. Find the values for Spark version and the Spark home directory. When you type the commands, run them either by pressing the Shift + Enter keys, or by clicking on the Play icon to the right of the word Ready.



NOTE:

The first time this is run, it may take a few minutes to complete. Future commands will run much faster, including this one if repeated.

```
sc.version  
sc.getConf.get("spark.home")
```

```
sc.version  
sc.getConf.get("spark.home")
```



While processing, Zeppelin will display a status of RUNNING. It will also display a Pause icon should it become necessary.

```
sc.version  
sc.getConf.get("spark.home")
```

RUNNING 0%

The output may vary slightly from the screenshot below, but should look something like this when processing is completed:

```
sc.version
sc.getConf.get("spark.home")
res0: String = 1.6.0
res1: String = /usr/hdp/2.4.0.0-169/spark
```

FINISHED ▶ ❌ 📄 ⚙️

- g. Zeppelin can be instructed to use multiple languages in an interactive fashion within the same notebook. Simply specify the desired language prior to the command.

Run the following commands to demonstrate this flexibility using Shell, Python, Scala, Markdown, and Spark SQL. Execute each command by clicking on the Play icon or pressing Shift + Enter when you are finished typing.

Shell:

```
%sh echo "Introduction to Zeppelin"
```

```
%sh echo "Introduction to Zeppelin"
Introduction to Zeppelin
```

Python:

```
%pyspark
print "Introduction to Zeppelin"
```

```
%pyspark
print "Introduction to Zeppelin"
Introduction to Zeppelin
```

Scala (default, so no need to specify prior to running command):

```
val s = "Introduction to Zeppelin"
```

```
val s = "Introduction to Zeppelin"
s: String = Introduction to Zeppelin
```

Markdown:

```
%md Introduction to Zeppelin
```

```
%md Introduction to Zeppelin
Introduction to Zeppelin
```

Spark SQL:

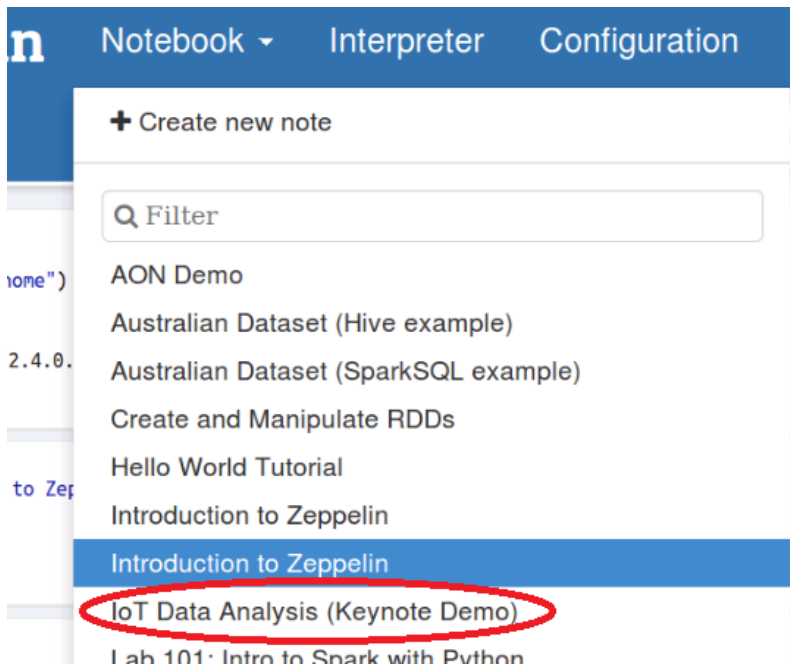
```
%sql  
show tables
```

The screenshot shows the Zeppelin SQL interface. At the top, there is a text input field containing the command '%sql show tables'. Below this is a toolbar with icons for table, bar chart, pie chart, area chart, line chart, and scatter plot. Below the toolbar is a table with two columns: 'tableName' and 'isTemporary'.

3. Use a preconfigured notebook to browse Zeppelin’s capabilities.

- a. Zeppelin has four major functions: data ingestion, data discovery, data analytics, and data visualization. One of the easiest ways to explore these functions is with a preconfigured notebook, many of which are available by default.

Click on Notebook at the top of the browser window and find and select the notebook labeled IoT Data Analysis (Keynote Demo) in the resulting drop-down menu.



Lab 2: Introduction to Spark REPLs and Zeppelin

b. At the top right click on the gear icon to change interpreter binding.



Introduction to Zeppelin ▶ ⏪ ⏩ ⏴ ⏵ ⏶ ⏷ ⏸ ⏹ ⏺ ⏻ ⏼ ⏽ ⏾ ⏿ ⏿ 🔍 🔒 🔒 default ▾

interpreter: vmr4nng

Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu.

- spark %spark (default), %spark, %sd, %dep
- md %md
- angular %angular
- sh %sh
- hive %hive
- tajo %tajo
- flink %flink
- lens %lens
- ignite %ignite, %ignite.igniteq
- cassandra %cassandra
- psql %psql
- phoenix %phoenix
- kylin %kylin

spark-yarn-client %spark, %spark.zyspark, %spark.sq, %spark.dep

Drag the spark-yarn-client to the top and click save.

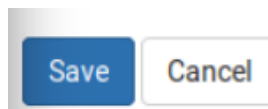
Introduction to Zeppelin ▶ ⏪ ⏩ ⏴ ⏵ ⏶ ⏷ ⏸ ⏹ ⏺ ⏻ ⏼ ⏽ ⏾ ⏿ ⏿ 🔍 🔒 🔒 default ▾

interpreter: vmr4nng

Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu.

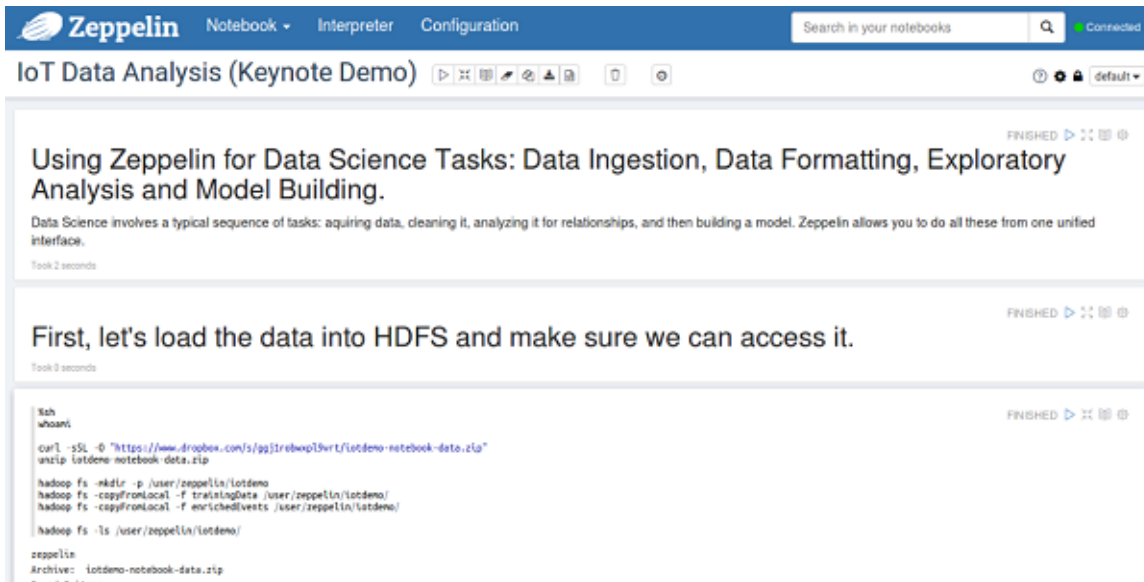
- spark-yarn-client %spark, %spark.zyspark, %spark.sq, %spark.dep
- spark %spark (default), %spark, %sd, %dep
- md %md
- angular %angular
- sh %sh
- hive %hive
- tajo %tajo
- flink %flink
- lens %lens
- ignite %ignite, %ignite.igniteq
- cassandra %cassandra
- psql %psql
- phoenix %phoenix
- kylin %kylin

The first interpreter on the list becomes default.



Lab 2: Introduction to Spark REPLs and Zeppelin

- c. For the purposes of this lab, all necessary code has already been entered for you in the saved notebook. All you have to do is scroll to the appropriate section and click the Play icon or press Shift + Enter.



The screenshot shows the Zeppelin notebook interface. At the top, there's a navigation bar with 'Zeppelin', 'Notebook', 'Interpreter', and 'Configuration'. A search bar and 'Connected' status are on the right. The notebook title is 'IoT Data Analysis (Keynote Demo)'. Below the title, there are three code blocks, each with a 'FINISHED' status and a play icon. The first block is a text introduction. The second block is another text introduction. The third block contains shell commands:

```
%sh
whoami

curl -sSL -O "https://www.dropbox.com/s/gg1robwxpl9vrt/iotdemo-notebook-data.zip"
unzip iotdemo-notebook-data.zip

hadoop fs -mkdir -p /user/zeppelin/iotdemo
hadoop fs -copyFromLocal -f trainingData /user/zeppelin/iotdemo/
hadoop fs -copyFromLocal -f enrichedEvents /user/zeppelin/iotdemo/

hadoop fs -ls /user/zeppelin/iotdemo/

zeppelin
Archive: iotdemo-notebook-data.zip
```

- d. The first major block of code ingests data from an online source into HDFS and then displays those files using the shell scripting interpreter. Find and run that code.



NOTE:

The label to the left of the Play icon says FINISHED, but this will not prohibit you from running the code again on this machine.

This notebook uses a deprecated command, `hadoop fs`, rather than the more updated `hdfs dfs` command we used in the previous lab. This should not affect the functionality of the demo.



The screenshot shows a code block in the Zeppelin notebook. The status is 'FINISHED' with a play icon. The code is:

```
%sh
whoami

curl -sSL -O "https://www.dropbox.com/s/gg1robwxpl9vrt/iotdemo-notebook-data.zip"
unzip iotdemo-notebook-data.zip

hadoop fs -mkdir -p /user/zeppelin/iotdemo
hadoop fs -copyFromLocal -f trainingData /user/zeppelin/iotdemo/
hadoop fs -copyFromLocal -f enrichedEvents /user/zeppelin/iotdemo/

hadoop fs -ls /user/zeppelin/iotdemo/
```

When the code has finished, the output at the bottom should look like this:

```

| hadoop fs -ls /user/zeppelin/iotdemo/
zeppelin
Archive: iotdemo-notebook-data.zip
Found 2 items
-rw-r--r--  3 zeppelin zeppelin      63570 2016-05-27 16:50 /user/zeppelin/iotdemo/enrichedEvents
-rw-r--r--  3 zeppelin zeppelin     33084 2016-05-27 16:50 /user/zeppelin/iotdemo/trainingData

```

- e. The next section of the notebook once again uses the shell scripting interpreter to view some of the raw data in one of the downloaded files. Scroll down and run this code, then view its output.

```

| %sh
| hadoop fs -cat /user/zeppelin/iotdemo/enrichedEvents | tail -n 10
Overspeed,"Y","hours",45,2773,-90.07,35.68,0,1,1
Lane Departure,"Y","hours",45,2773,-90.04,35.19,1,1,0
Normal,"Y","hours",45,2773,-90.68,35.12,1,0,0
Normal,"Y","hours",45,2773,-91.14,34.96,0,0,0
Normal,"Y","hours",45,2773,-91.93,34.81,0,0,0
Normal,"Y","hours",45,2773,-92.31,34.78,0,1,0
Normal,"Y","hours",45,2773,-92.09,34.8,0,0,0
Normal,"Y","hours",45,2773,-91.93,34.81,0,0,0
Normal,"Y","hours",45,2773,-90.68,35.12,1,0,0
Normal,"Y","hours",45,2773,-91.14,34.96,0,0,0

```

- f. The next section of the notebook performs actions necessary to import and use this data with Spark SQL. You may note that the status to the left of the Play icon is shown as ERROR. This is due to the fact that the file being manipulated did not exist at the time the notebook was opened on this system. Run this code and view the output.

```

val sqlContext = new org.apache.spark.sql.SQLContext(sc)

val eventsFile = sc.textFile("hdfs:///user/zeppelin/iotdemo/enrichedEvents")

case class Event(eventType: String,
                 isCertified: String,
                 paymentScheme: String,
                 hoursDriven: Int,
                 milesDriven: Int,
                 lat: Float,
                 long: Float,
                 isFoggy: Int,
                 isRainy: Int)

```

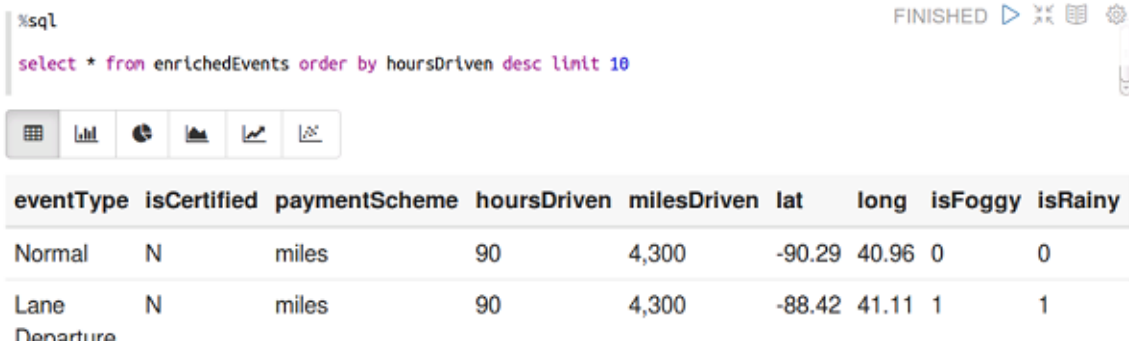
Lab 2: Introduction to Spark REPLs and Zeppelin

The output should look like this:

```
eventsRDD.toDF().registerTempTable("enrichedEvents")
```

```
sqlContext: org.apache.spark.sql.SQLContext = org.apache.spark.sql.SQLContext@312d2a12
eventsFile: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[3] at textFile at <console>:29
defined class Event
eventsRDD: org.apache.spark.rdd.RDD[Event] = MapPartitionsRDD[5] at map at <console>:35
res4: Long = 1359
```

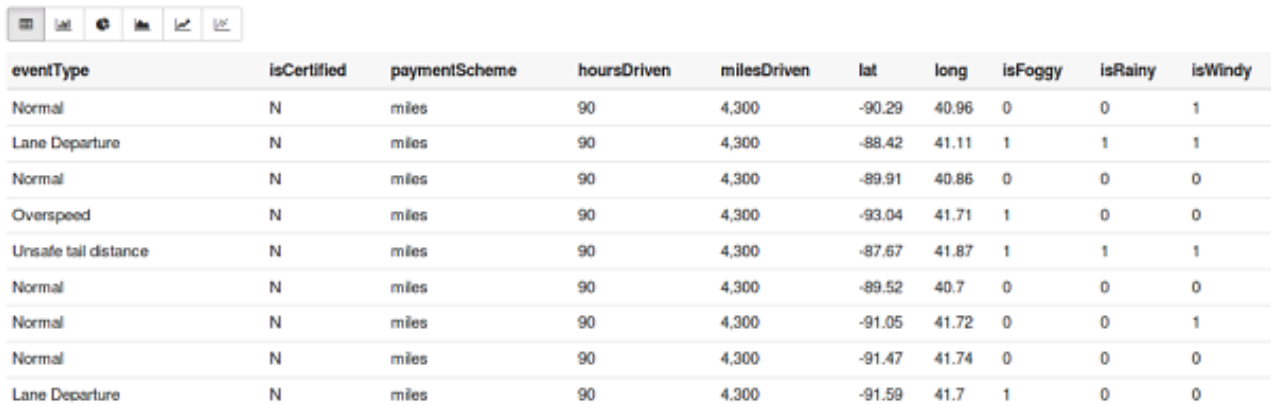
- g. The next block of code utilizes Spark SQL to view this data. Run this code and examine the output.



The screenshot shows a Zeppelin notebook interface. At the top, there is a code editor with the following Spark SQL query: `%sql select * from enrichedEvents order by hoursDriven desc limit 10`. The status bar indicates "FINISHED". Below the code editor, there is a toolbar with six visualization icons: a table icon (selected), a bar chart, a pie chart, a line chart, a scatter plot, and a heatmap. The output of the query is displayed as a table with the following data:

eventType	isCertified	paymentScheme	hoursDriven	milesDriven	lat	long	isFoggy	isRainy
Normal	N	miles	90	4,300	-90.29	40.96	0	0
Lane Departure	N	miles	90	4,300	-88.42	41.11	1	1

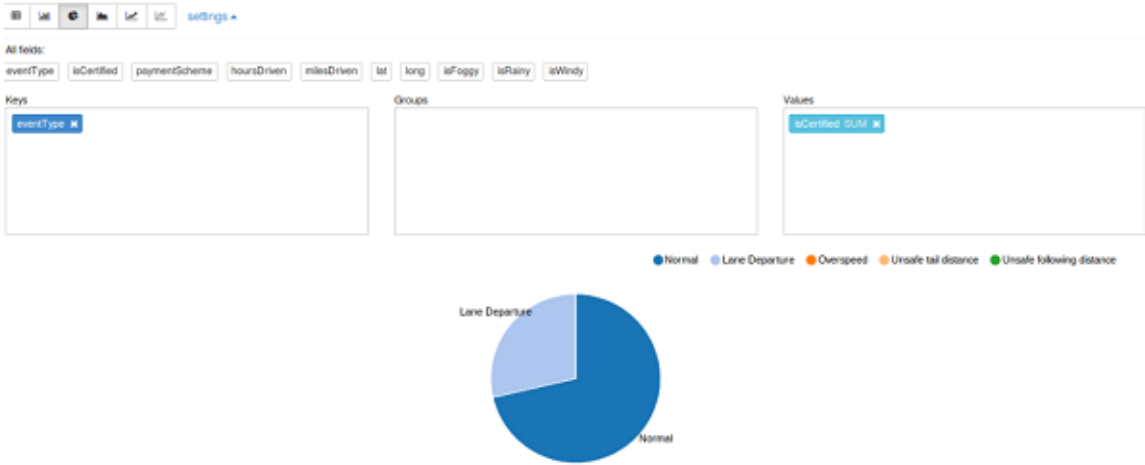
- h. Note that at the top of the results there are six buttons that allow you to display the results using six different visualizations. Click on each one to view the differences between them.



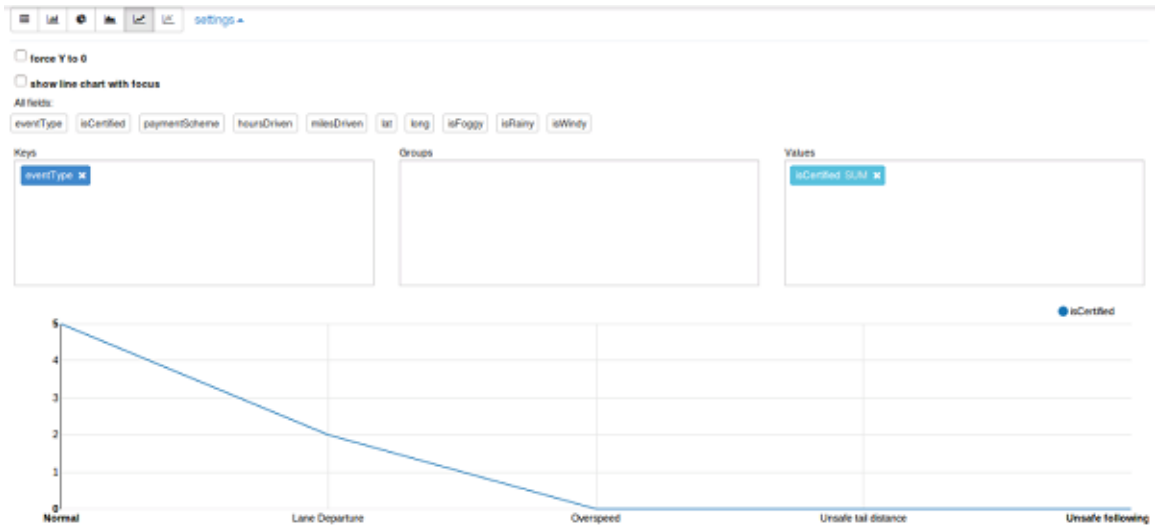
The screenshot shows the same Zeppelin notebook interface, but with the table visualization selected in the toolbar. The output is a table with the following data:

eventType	isCertified	paymentScheme	hoursDriven	milesDriven	lat	long	isFoggy	isRainy	isWindy
Normal	N	miles	90	4,300	-90.29	40.96	0	0	1
Lane Departure	N	miles	90	4,300	-88.42	41.11	1	1	1
Normal	N	miles	90	4,300	-89.91	40.86	0	0	0
Overspeed	N	miles	90	4,300	-93.04	41.71	1	0	0
Unsafe tail distance	N	miles	90	4,300	-87.67	41.87	1	1	1
Normal	N	miles	90	4,300	-89.52	40.7	0	0	0
Normal	N	miles	90	4,300	-91.05	41.72	0	0	1
Normal	N	miles	90	4,300	-91.47	41.74	0	0	0
Lane Departure	N	miles	90	4,300	-91.59	41.7	1	0	0

Lab 2: Introduction to Spark REPLs and Zeppelin



Lab 2: Introduction to Spark REPLs and Zeppelin



TIP:

In this lab you ran each section of code, known as a paragraph, individually. The entire notebook could have been played at once, however, by clicking the Play icon labeled Run all paragraphs directly to the right of the notebook title at the top of the browser.

IoT Data Analysis (Keynote Demo)



Result

You have accessed the Spark REPLs for both Scala and Python, created a Zeppelin notebook and demonstrated Zeppelin's ability to interpret multiple languages, and used a pre-built Zeppelin notebook to briefly explore Zeppelin's ability to ingest, view, analyze, and visualize data.

Lab 3: Creating and Manipulating RDDs (Scala)

About This Lab

Objective:

Create and Manipulate RDDs using Scala and Zeppelin

File Locations:

/home/zeppelin/spark/data/

Successful Outcome:

Perform basic RDD transformations and actions using Zeppelin.

Before You Begin:

Complete the Pre-Lab

Lab Steps

Perform the following steps:

1. View the raw data for this lab.

- a. In a new terminal window, ssh to sandbox and change directories to /home/zeppelin/spark/data. View the files in this directory.

```
# ssh sandbox
# cd /home/zeppelin/spark/data/
# ls
```

```
root@ubuntu:~# ssh sandbox
Last login: Mon May 30 09:23:47 2016 from ip-172-17-42-1.ec2.internal
[root@sandbox ~]# cd /home/zeppelin/spark/data
[root@sandbox data]# ls
airports.csv  data.txt      plane-data.csv  small_blocks.txt
carriers.csv  flights.csv   selfishgiant.txt
[root@sandbox data]#
```

- b. Use `less` to view the “selfishgiant.txt” data file. Press `q` to quit when you are finished reviewing.

```
# less selfishgiant.txt
```

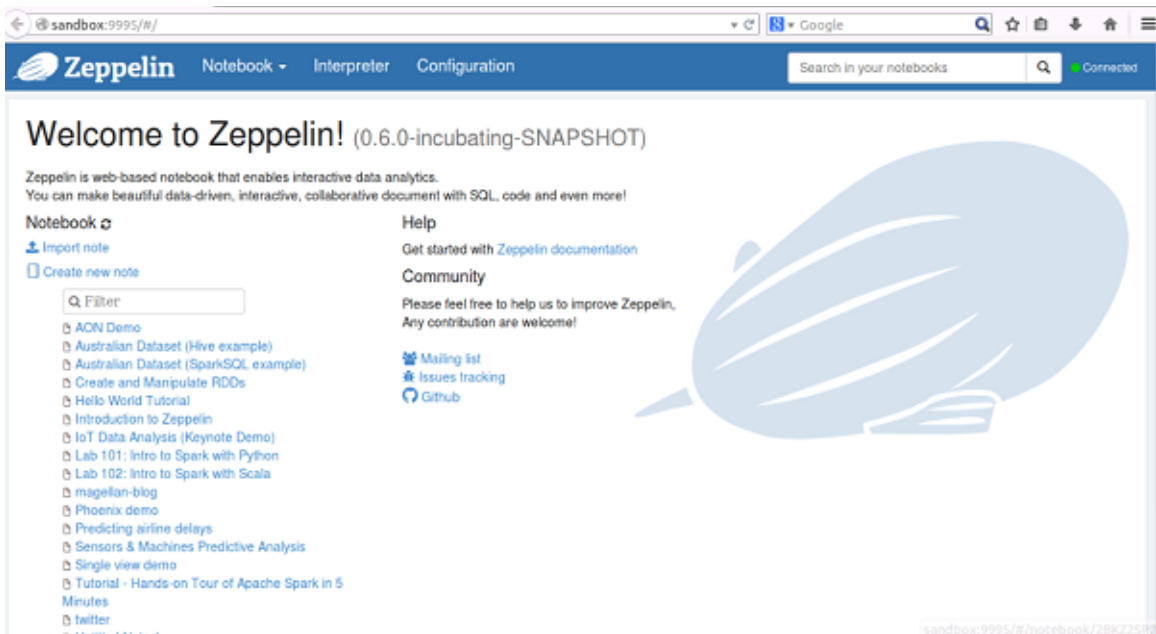
```
[root@sandbox data]# less selfishgiant.txt
```

Lab 3: Creating and Manipulating RDDs (Scala)

```
EVERY afternoon, as they were coming from school, the children used to go and play in the Giant's garden.^^^It was a large lovely garden, with soft green grass. Here and there over the grass stood beautiful flowers like stars, and there were twelve peach-trees that in the springtime broke out into delicate blossoms of pink and pearl, and in the autumn bore rich fruit. The birds sat on the trees and sang so sweetly that the children used to stop their games in order to listen to them. <D4>How happy we are here!<D5> they cried to each other.^^^One day the Giant came back. He had been to visit his friend the Cornish ogre, and had stayed with him for seven years. After the seven years were over he had said all that he had to say, for his conversation was limited, and he determined to return to his own castle. When he arrived he saw the children playing in the garden.^^^<D4>What are you doing there?<D5> he cried in a very gruff voice, and the children ran away.^^^<D4>My own garden is my own garden,<D5> said the Giant; <D4>and any one can understand that, and I will allow nobody to play in it but myself.<D5> So he built a high wall all round it, and put up a notice-board.^^^TRESPASSERS ^^WILL BE ^^PROSECUTED^^He was a very selfish Giant.^^^The poor children had now nowhere to play. They tried to play on the road, but the road was very dusty and full of hard stones, and they did not like it. They used to wander round the high wall when their lessons were over, and talk about the beautiful garden inside. <D4>How happy we were there,<D5> they said to each other.^^^Then the Spring came, and all over the country there were little blossoms and little birds. Only in the garden of the Selfish Giant it was still winter. The birds did not care to sing in it as there were no children, and the trees forgot to blossom.:
```

2. Perform basic RDD manipulations using the Zeppelin notebook.

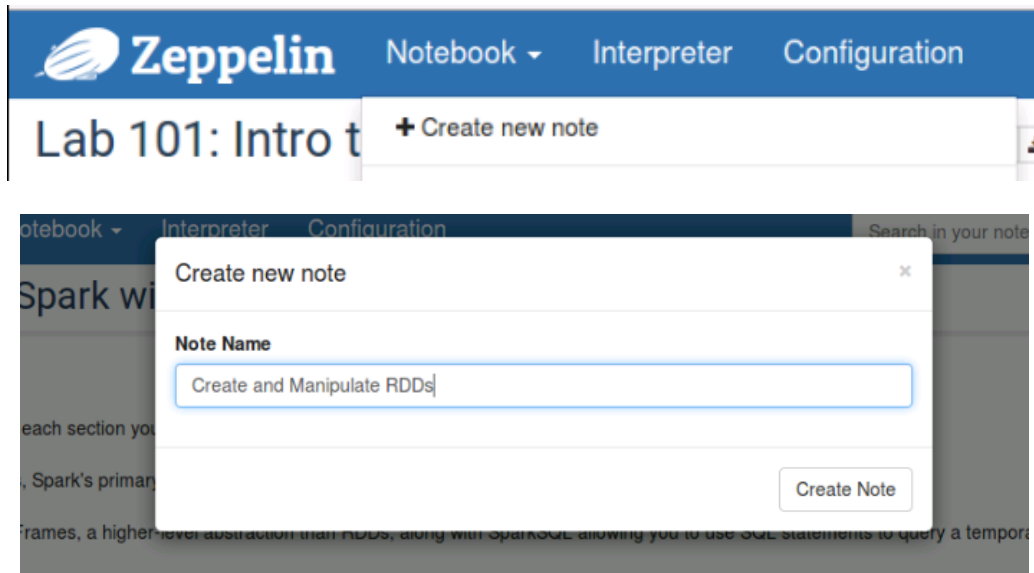
- Open the Firefox browser and enter the following URL to view the Zeppelin UI.
<http://sandbox:9995/>



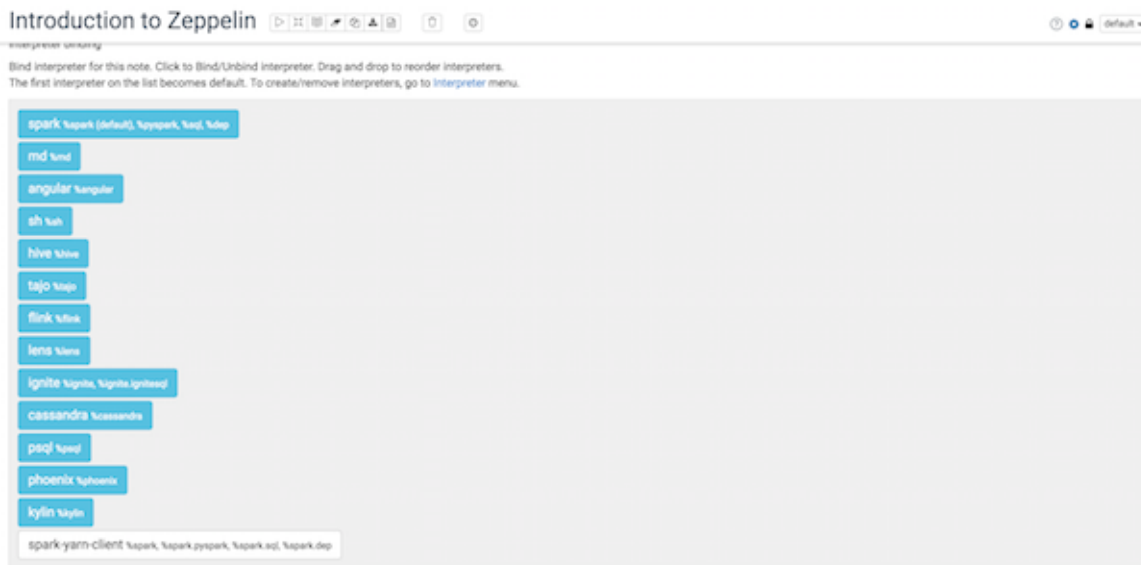
The screenshot shows the Zeppelin notebook interface in a browser. The page title is "Welcome to Zeppelin! (0.6.0-incubating-SNAPSHOT)". Below the title, there is a brief description of Zeppelin as a web-based notebook for interactive data analytics. The main content area is divided into two columns. The left column contains a "Notebook" section with a search filter and a list of notebooks, including "Create and Manipulate RDDs". The right column contains a "Help" section with links to documentation, a "Community" section with a message about contributing, and links to "Mailing list", "Issues tracking", and "Github". A large blue zeppelin logo is visible on the right side of the page.

Lab 3: Creating and Manipulating RDDs (Scala)

- b. Click on Notebook and select Create new note on the drop down. Name this note Create and Manipulate RDDs.

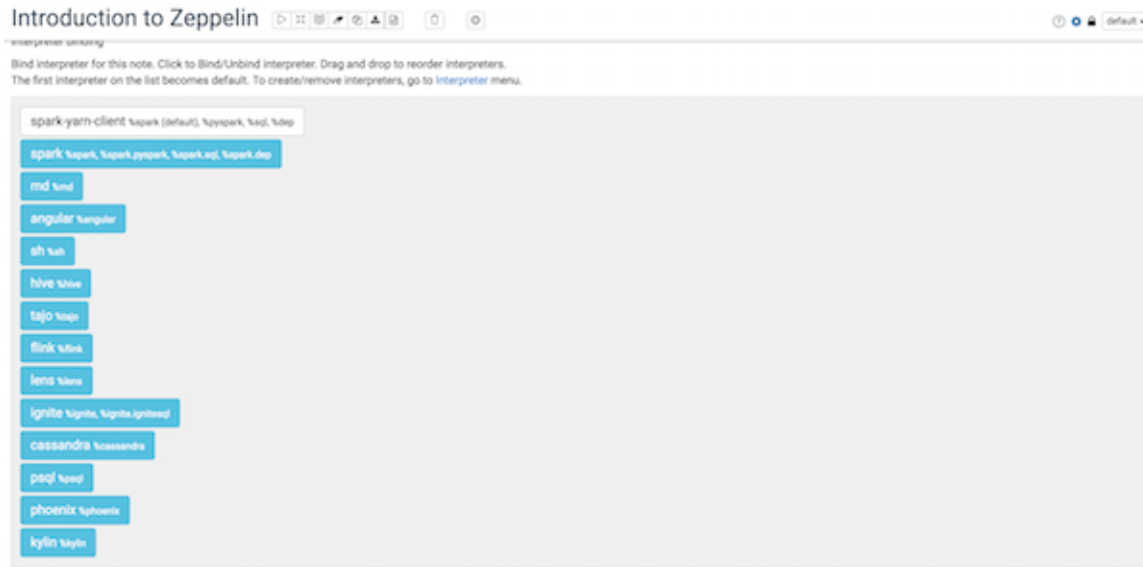


- c. At the top right click on the gear icon to change interpreter binding.

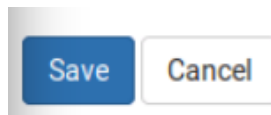


Lab 3: Creating and Manipulating RDDs (Scala)

Drag the spark-yarn-client to the top and click save.



The first interpreter on the list becomes default.



- d. Place the selfishgiant.txt file into the Zeppelin user's home directory on HDFS, /user/zeppelin. (There are no line breaks in the code below after %sh. Please refer to the screenshot.)

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/selfishgiant.txt
/user/zeppelin/selfishgiant.txt
```

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/selfishgiant.txt /user/zeppelin/selfishgiant.txt
```



REMINDER:

After entering a command, press **Shift + Enter** keys or press the **Play** button on the right side of the paragraph to execute the commands. The text to the left of the **Play** button should change from **READY** to **FINISHED** when it is complete.

FINISHED    

- e. Verify the file was uploaded successfully.

```
%sh
hdfs dfs -ls /user/zeppelin
```

```
%sh
hdfs dfs -ls /user/zeppelin

Found 5 items
drwx----- - zeppelin zeppelin          0 2016-04-25 20:00 /user/zeppelin/.Trash
drwxr-xr-x - zeppelin zeppelin          0 2016-05-27 16:06 /user/zeppelin/.sparkStaging
-rw-r--r-- 3 zeppelin zeppelin    4610348 2016-04-18 09:24 /user/zeppelin/bank-full.csv
drwxr-xr-x - zeppelin zeppelin          0 2016-05-27 16:50 /user/zeppelin/iotdemo
-rw-r--r-- 3 zeppelin zeppelin      8596 2016-05-30 10:52 /user/zeppelin/selfishgiant.txt
```

- f. Create an RDD named `baseRdd` using this file. Verify the RDD exists by using the `take()` function to print the first line of the file.

```
val baseRdd = sc.textFile("/user/zeppelin/selfishgiant.txt")
baseRdd.take(1)
```

```
val baseRdd = sc.textFile("/user/zeppelin/selfishgiant.txt")
baseRdd.take(1)

baseRdd: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[32] at textFile at <console>:29
res10: Array[String] = Array(EVERY afternoon, as they were coming from school, the children used to go and play in t
he Giant's garden.)
```

- g. Each line of the file is currently a string. Transform the lines into arrays of individual elements (words) stored in a new RDD named `splitRdd`, then take a look at the first five elements.

```
val splitRdd = baseRdd.flatMap(line => line.split(" "))
splitRdd.take(5)
```

```
val splitRdd = baseRdd.flatMap(line => line.split(" "))
splitRdd.take(5)

splitRdd: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[33] at flatMap at <console>:31
res12: Array[String] = Array(EVERY, afternoon,, as, they, were)
```

- h. Create a new RDD named `filterRdd` that only contains words in `splitRdd` that are longer than 10 characters. Use `collect()` to view the entire output.

```
val filterRdd = splitRdd.filter(word => word.length() > 10)
filterRdd.collect()
```

- i. Display a count of the total number of words in `splitRdd`.

```
splitRdd.count()
```

```
splitRdd.count()
res16: Long = 1683
```

- j. Create an RDD named `distinctRdd` that eliminates any duplicate words in `splitRdd`. Then display a count of the number of distinct words in the RDD.

```
val distinctRdd = splitRdd.distinct()
distinctRdd.count()
```

```
val distinctRdd = splitRdd.distinct()
distinctRdd.count()
distinctRdd: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[43] at distinct at <console>:33
res20: Long = 594
```

- k. Save the contents of `distinctRDD` to text in HDFS. Put the contents in a folder named “distinct” for future reference.

```
distinctRdd.saveAsTextFile("/user/zeppelin/distinct")
```

```
distinctRdd.saveAsTextFile("/user/zeppelin/distinct")
```

- l. Verify the contents of the RDD were written to HDFS.

```
%sh
hdfs dfs -ls /user/zeppelin/distinct
```

```
%sh
hdfs dfs -ls /user/zeppelin/distinct
Found 3 items
-rw-r--r--  3 zeppelin zeppelin      0 2016-05-30 11:37 /user/zeppelin/distinct/_SUCCESS
-rw-r--r--  3 zeppelin zeppelin    1987 2016-05-30 11:37 /user/zeppelin/distinct/part-00000
-rw-r--r--  3 zeppelin zeppelin    1860 2016-05-30 11:37 /user/zeppelin/distinct/part-00001
```

- m. View the contents of one of the `part-*` files and verify that an array of unique words has been generated and saved. (Your output may differ from the screenshot below.)

```
%sh
hdfs dfs -cat /user/zeppelin/distinct/part-00001
```

Lab 3: Creating and Manipulating RDDs (Scala)

```
%sh
hdfs dfs -cat /user/zeppelin/distinct/part-00001

furs,
BE
since
Winter
don't
spot,
wall,
felt
wall.
seen
tree,
tree.
away.
covered
corner
still
children
```

- n. Create an RDD named `numbersRdd` that contains an array of the following numbers: 15, 20, 95, and 80. View the contents of the RDD to verify it was successfully created.

```
val numbersRdd = sc.parallelize(List(15, 20, 95, 80))
numbersRdd.collect()
```

```
val numbersRdd = sc.parallelize(List(15, 20, 95, 80))
numbersRdd.collect()
FINISHED

numbersRdd: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[46] at parallelize at <console>:29
res25: Array[Int] = Array(15, 20, 95, 80)
```

- o. Display a count of the elements in `numbersRdd`, as well as the mean, standard deviation, maximum, and minimum values.

```
numbersRdd.stats()
```

```
numbersRdd.stats()
FINISHED

res27: org.apache.spark.util.StatCounter = (count: 4, mean: 52.500000, stdev: 35.443617, max: 95.000000, min: 15.000000)
```

Lab 3: Creating and Manipulating RDDs (Scala)

- p. Create a variable named `maryFile` that contains the string “Mary had a little lamb” and then convert that variable into an RDD named `maryRdd`. View the RDD contents when finished.

```
val maryFile = Array("Mary had a little lamb")
val maryRdd = sc.parallelize(maryFile)
maryRdd.collect()
```

```
val maryFile = Array("Mary had a little lamb")
val maryRdd = sc.parallelize(maryFile)
maryRdd.collect()
```

FINISHED

```
maryFile: Array[String] = Array(Mary had a little lamb)
maryRdd: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[49] at parallelize at <console>:31
res30: Array[String] = Array(Mary had a little lamb)
```

- q. In Scala, to combine RDDs their types must match. Thus, if we attempt to perform a `union()` using `maryRdd` and `numbersRdd` at this point, we will get a type mismatch error. To alleviate this before the next lab step, cast the integer values in `numbersRdd` to strings in an RDD named `numbersRddString`.

```
val numbersRddString = numbersRdd.map(num => num.toString)
```

```
val numbersRddString = numbersRdd.map(num => num.toString)
numbersRddString.collect()
```

FINISHED

```
numbersRddString: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[52] at map at <console>:31
res34: Array[String] = Array(15, 20, 95, 80)
```



NOTE:

In Python, casting the integers to string values before performing a `union()` is handled automatically, thus the equivalent to this step does not exist in the Python version of the lab book.

Lab 3: Creating and Manipulating RDDs (Scala)

- r. Create a new RDD named `comboRdd` that creates a union between `maryRdd` and `numbersRddString`. Then view the combined RDD.

```
val comboRdd = maryRdd.union(numbersRddString)
comboRdd.collect()
```

```
val comboRdd = maryRdd.union(numbersRddString)
comboRdd.collect()
```

```
comboRdd: org.apache.spark.rdd.RDD[String] = UnionRDD[53] at union at <console>:37
res36: Array[String] = Array(Mary had a little lamb, 15, 20, 95, 80)
```

Result

You have created several RDDs and performed various transactions and actions using the Zeppelin notebook.

Lab 4: Create and Manipulate Pair RDDs (Scala)

About This Lab

Objective:

Create pair RDD's and use various functions to transform these RDD's using Scala in Zeppelin.

File Locations:

/home/zeppelin/spark/data/

Successful Outcome:

REQUIRED: Create pair RDDs and perform various operations.

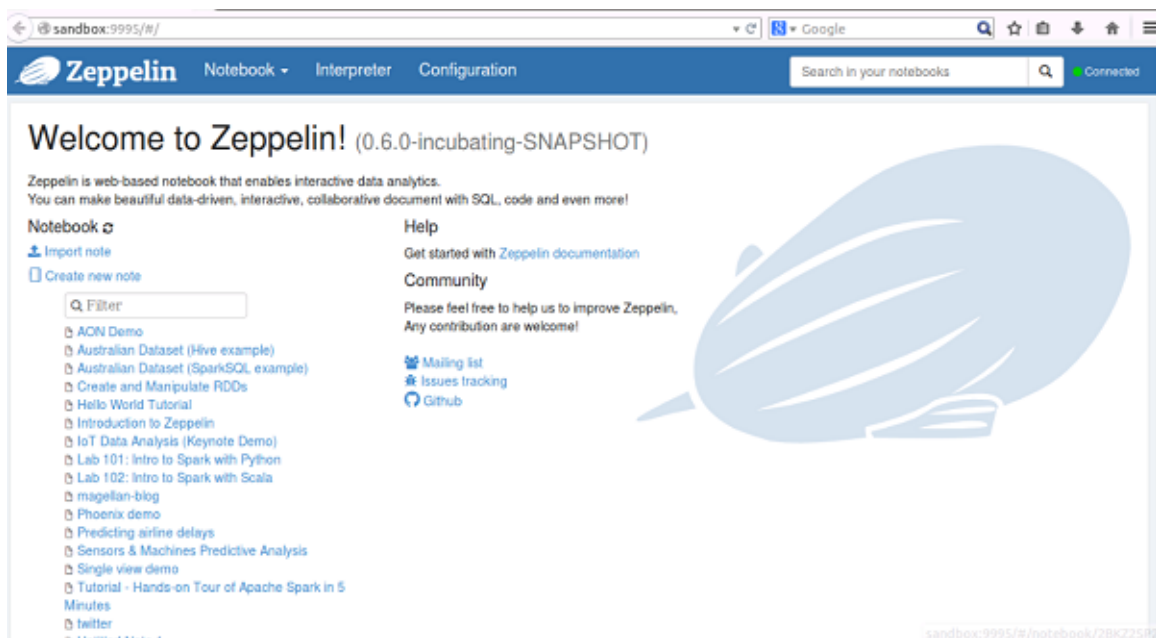
OPTIONAL: Complete challenge labs performing more complex operations.

Lab Steps

Perform the following steps:

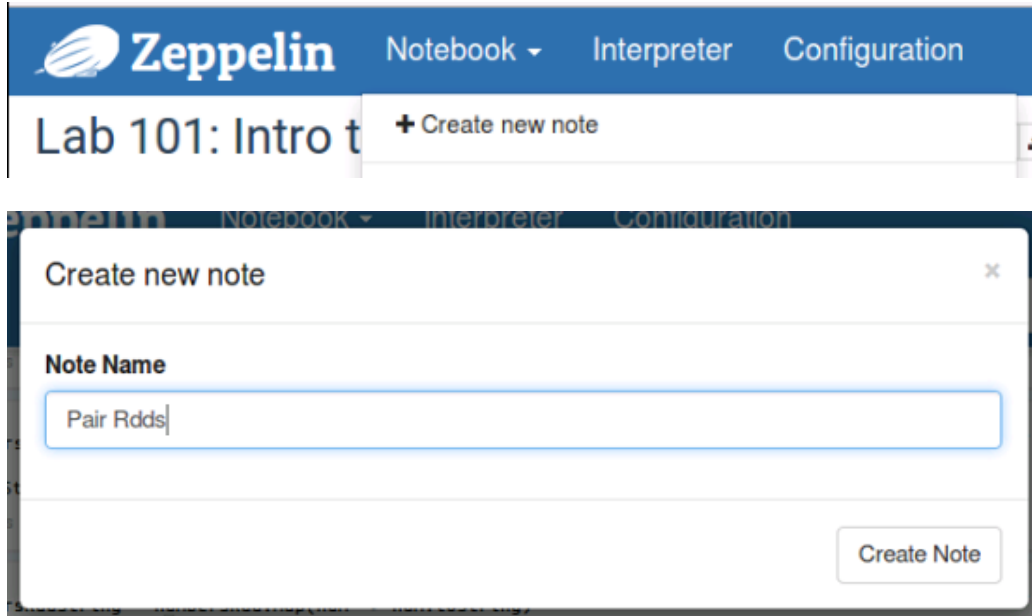
1. Create a Pair RDD note in Zeppelin.

- a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.
<http://sandbox:9995/>

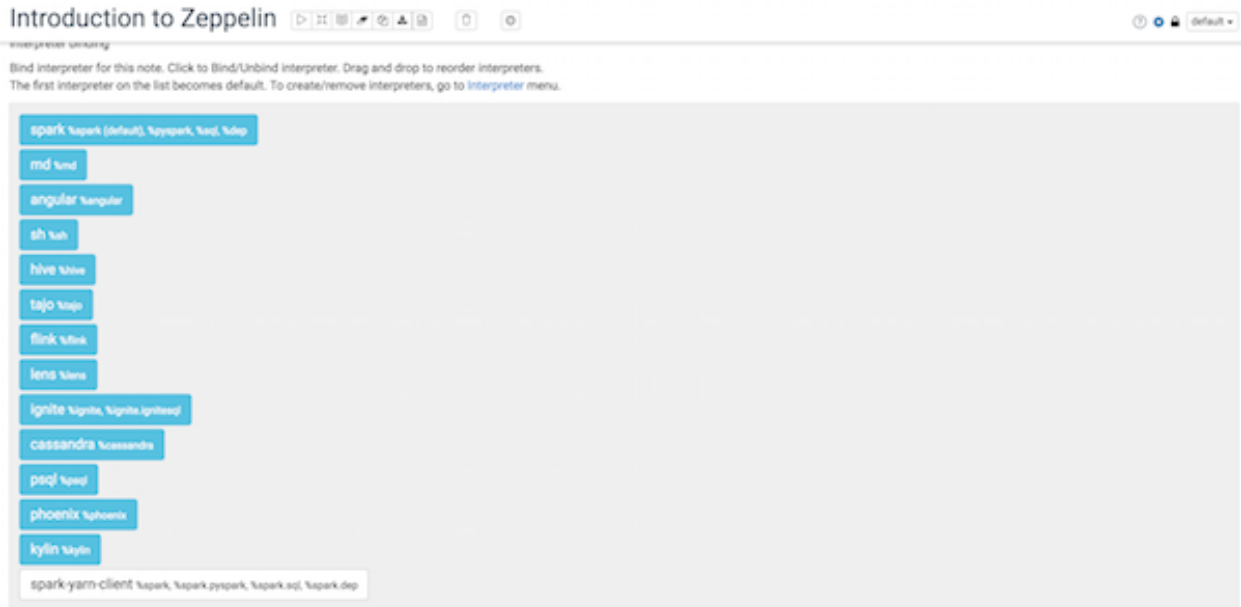


Lab 4: Create and Manipulate Pair RDDs (Scala)

- b. Click on Notebook and select Create new note on the drop down. Name this note Pair RDDs.

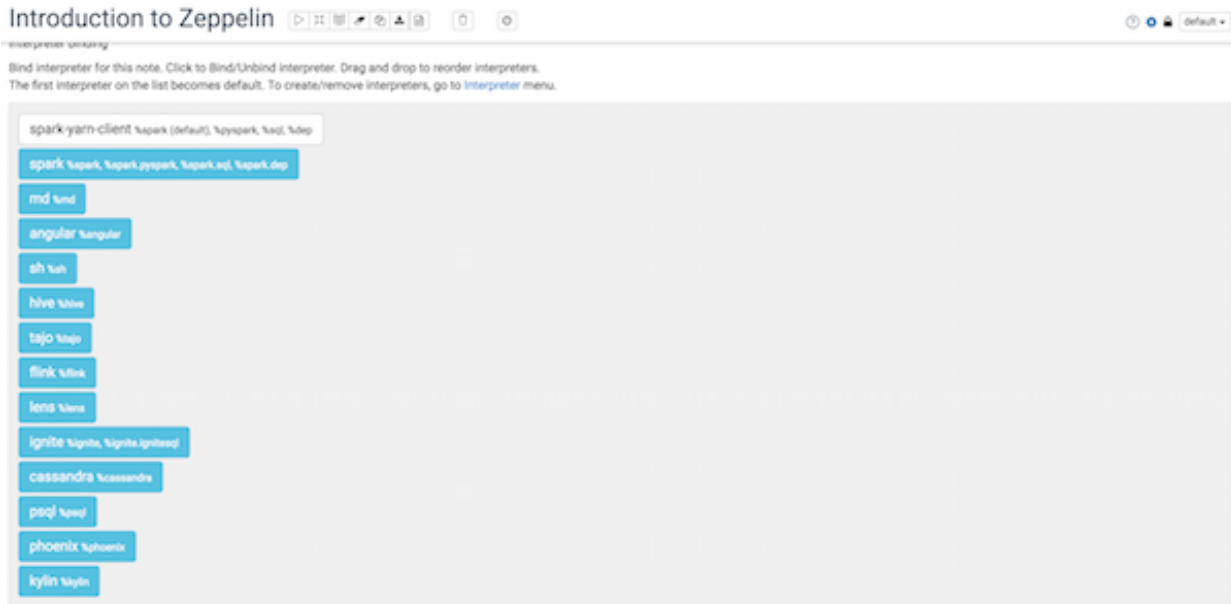


- c. At the top right click on the gear icon to change interpreter binding.

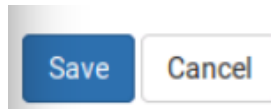


Lab 4: Create and Manipulate Pair RDDs (Scala)

Drag the spark-yarn-client to the top and click save.



The first interpreter on the list becomes default.



2. Create a Pair RDD from a text file using map().

- Recreate the RDD `splitRDD` using the `selfishgiant.txt` file by importing it to an RDD as a text file and then flattening it into individual word elements. Then view the first 5 words to confirm the RDD exists and is correctly formatted.

In the code below, there are no line breaks between `splitRdd` and `(" ")`. Please refer to the screenshot.

```
val splitRdd = sc.textFile("/user/zeppelin/selfishgiant.txt").flatMap(line =>
line.split(" "))

splitRdd.take(5)
```

```
val splitRdd = sc.textFile("/user/zeppelin/selfishgiant.txt").flatMap(line => line.split(" "))
splitRdd.take(5)

splitRdd: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[226] at flatMap at <console>:29
res40: Array[String] = Array(EVERY, afternoon,, as, they, were)
```

**NOTE:**

In the previous lab, this RDD creation was performed over two steps, creating an intermediary RDD named `baseRdd`. The creation of the intermediary is not necessary unless it needs to be used in a future step.

- b. Use `map()` to create an RDD named `mappedRdd` that converts each element into a key-value pair with a value of 1. View the first five elements to confirm successful operation.

```
val mappedRdd = splitRdd.map(word => (word, 1))
mappedRdd.take(5)
```

```
val mappedRdd = splitRdd.map(word => (word, 1))
mappedRdd.take(5)
```

```
mappedRdd: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[227] at map at <console>:31
res42: Array[(String, Int)] = Array((EVERY,1), (afternoon,,1), (as,1), (they,1), (were,1))
```

3. Create Pair RDDs using zip functions and perform simple transformations.

- a. Create a variable named `months` that contains the values `Jan`, `Feb`, `Mar`, `Apr`, `May`, `Jun`, and `Jul` as a list of string values. Convert this to an RDD named `monthsRdd`. Then create another RDD named `monthsIndexed0Rdd` using `zipWithIndex()` to create a Pair RDD that automatically assigns a value to each element based on its position in the list.

**REMINDER:**

The first element will be assigned a value of “0” using this function.

```
val months = Array("Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul")
val monthsRdd = sc.parallelize(months)
val monthsIndexed0Rdd = monthsRdd.zipWithIndex()
monthsIndexed0Rdd.collect()
```

```
val months = Array("Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul")
val monthsRdd = sc.parallelize(months)
val monthsIndexed0Rdd = monthsRdd.zipWithIndex()
monthsIndexed0Rdd.collect()
```

FINISHED ▶ ⌂ ☰

```
months: Array[String] = Array(Jan, Feb, Mar, Apr, May, Jun, Jul)
monthsRdd: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[228] at parallelize at <console>:31
monthsIndexed0Rdd: org.apache.spark.rdd.RDD[(String, Long)] = ZippedWithIndexRDD[229] at zipWithIndex at <console>:33
res44: Array[(String, Long)] = Array((Jan,0), (Feb,1), (Mar,2), (Apr,3), (May,4), (Jun,5), (Jul,6))
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

- b. Use `map()` to convert the value for each month to the actual month number and store this in a new RDD named `monthsIndexed1Rdd`. For reference, Jan should have a value of 1, Feb should have a value of 2, and so on. View the new RDD to confirm success.

```
val monthsIndexed1Rdd = monthsIndexed0Rdd.map{case (x,y) => (x, y + 1)}  
monthsIndexed1Rdd.collect()
```

```
val monthsIndexed1Rdd = monthsIndexed0Rdd.map{case (x,y) => (x, y+1)}  
monthsIndexed1Rdd.collect() FINISHED  
  
monthsIndexed1Rdd: org.apache.spark.rdd.RDD[(String, Long)] = MapPartitionsRDD[243] at map at <console>:35  
res67: Array[(String, Long)] = Array((Jan,1), (Feb,2), (Mar,3), (Apr,4), (May,5), (Jun,6), (Jul,7))
```

- c. Use `mapValues()` to convert the value for each month to the actual month number and store this in a new RDD named `monthsIndexed2Rdd`. For reference, Jan should have a value of 1, Feb should have a value of 2, and so on. View the new RDD to confirm success.

```
val monthsIndexed2Rdd = monthsIndexed0Rdd.mapValues(y => y + 1)  
monthsIndexed2Rdd.collect()
```

```
val monthsIndexed2Rdd = monthsIndexed0Rdd.mapValues(y => y + 1)  
monthsIndexed2Rdd.collect() FINISHED ▶ ;  
  
monthsIndexed2Rdd: org.apache.spark.rdd.RDD[(String, Long)] = MapPartitionsRDD[244] at mapValues at <console>:35  
res69: Array[(String, Long)] = Array((Jan,1), (Feb,2), (Mar,3), (Apr,4), (May,5), (Jun,6), (Jul,7))
```



NOTE:

No difference exists between the two previous lab steps from Spark's perspective. The `mapValues` function simply performs a `map()` and returns the key without modification, while performing the function you define on the value.

- d. Create a variable named `quarters` that contains the following seven values: 1, 1, 1, 2, 2, 2, and 3. Convert the variable into an RDD named `quartersRdd`. Then create an RDD named `monthsZipQuarters` and use `zip()` to create a Pair RDD that assigns each value from `quartersRdd` to a month in `monthsRdd`. Finally, view the output and make sure that each month was assigned to the correct quarter in the final RDD.

```
val quarters = Array(1, 1, 1, 2, 2, 2, 3)  
val quartersRdd = sc.parallelize(quarters)  
val monthsZipQuarters = monthsRdd.zip(quartersRdd)  
monthsZipQuarters.collect()
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

```
val quarters = Array(1, 1, 1, 2, 2, 2, 3)
val quartersRdd = sc.parallelize(quarters)
val monthsZipQuarters = monthsRdd.zip(quartersRdd)
monthsZipQuarters.collect()

quarters: Array[Int] = Array(1, 1, 1, 2, 2, 2, 3)
quartersRdd: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[233] at parallelize at <console>:31
monthsZipQuarters: org.apache.spark.rdd.RDD[(String, Int)] = ZippedPartitionsRDD2[234] at zip at <console>:37
res52: Array[(String, Int)] = Array((Jan,1), (Feb,1), (Mar,1), (Apr,2), (May,2), (Jun,2), (Jul,3))
```

FINISHED [

- e. Perform the following operations on `monthsZipQuarters` without creating new RDDs: view the keys only, view the values only, and view the contents of the RDD sorted alphabetically by key.

```
monthsZipQuarters.keys.collect()
monthsZipQuarters.values.collect()
monthsZipQuarters.sortByKey().collect()
```

```
monthsZipQuarters.keys.collect()
monthsZipQuarters.values.collect()
monthsZipQuarters.sortByKey().collect()

res59: Array[String] = Array(Jan, Feb, Mar, Apr, May, Jun, Jul)
res60: Array[Int] = Array(1, 1, 1, 2, 2, 2, 3)
res61: Array[(String, Int)] = Array((Apr,2), (Feb,1), (Jan,1), (Jul,3), (Jun,2), (Mar,1), (May,2))
```

4. Count the number of times words appear in a Pair RDD and manipulate the output.

- a. Use the `mappedRdd` created in a previous step and create a new RDD named `reducedByKeyRdd` that reduces the file so that each word appears only once but has a value equal to the number of times it appeared in the original RDD. View the first five elements of the new RDD to confirm successful operation.

```
val reducedByKeyRdd = mappedRdd.reduceByKey((x,y) => x+y)
reducedByKeyRdd.take(5)
```

```
val reducedByKeyRdd = mappedRdd.reduceByKey((x,y) => x+y)
reducedByKeyRdd.take(5)

reducedByKeyRdd: org.apache.spark.rdd.RDD[(String, Int)] = ShuffledRDD[242] at reduceByKey at <console>:33
res65: Array[(String, Int)] = Array((branches,4), (sweet,1), (here!,1), (rubbed,1), (country,1))
```

FINISHED [

Lab 4: Create and Manipulate Pair RDDs (Scala)

- b. Use `map()` to create a new RDD named `flippedRdd` that switches your keys and values so that the current keys become the values, and the values become the keys. View the first five elements of the new RDD to confirm successful operation.

```
val flippedRdd = reducedByKeyRdd.map{case (x,y) => (y,x)}  
flippedRdd.take(5)
```

```
val flippedRdd = reducedByKeyRdd.map{case (x,y) => (y,x)}  
flippedRdd.take(5) F  
flippedRdd: org.apache.spark.rdd.RDD[(Int, String)] = MapPartitionsRDD[245] at map at <console>:35  
res71: Array[(Int, String)] = Array((1,rubbed), (4,branches), (1,here!+), (1,under), (1,cold))
```

- c. Create a new RDD named `orderedRdd` that manipulates `flippedRDD` and arranges the words in descending order by number of times they appear. View the first five elements of the new RDD to confirm successful operation.

```
val orderedRdd = flippedRdd.sortByKey(ascending = false)  
orderedRdd.take(5)
```

```
val orderedRdd = flippedRdd.sortByKey(ascending = false)  
orderedRdd.take(5) FI  
orderedRdd: org.apache.spark.rdd.RDD[(Int, String)] = ShuffledRDD[248] at sortByKey at <console>:37  
res73: Array[(Int, String)] = Array((148,the), (85,and), (44,he), (38,to), (32,was))
```

Result

You have successfully created and manipulated Pair RDD's using various functions.

Challenge Labs

The labs below work with Pair RDDs to perform real-world operations. In some cases, the solutions to the lab utilize programming techniques not explicitly described in the course lecture. These techniques, however, should be clear and easy to understand by carefully following the instructions. If you have questions and are in an instructor-supported class, please ask for assistance as needed.

You may want to start by creating a new notebook named Pair RDD Challenge Labs, but this is up to you.

Perform the following steps:

1. Determine the airlines with the greatest number of flights.

- a. Go back to a terminal window that has used SSH to connect to the sandbox Docker environment and change to the `/home/zeppelin/spark/data` directory if necessary. View the contents of this directory and confirm the existence of three files: `airports.csv`, `plane-data.csv`, and `flights.csv`.

```
# ls
```

```
[zeppelin@sandbox data]# pwd
/home/zeppelin/spark/data
[zeppelin@sandbox data]# ls
airports.csv  data.txt      plane-data.csv  small_blocks.txt
carriers.csv  flights.csv   selfishgiant.txt
[zeppelin@sandbox data]#
```

- b. Use `head` to view the first few lines of the `flights.csv` file.

```
# head flights.csv
```

```
[zeppelin@sandbox data]# head flights.csv
1,3,4,2003,2211,WN,335,N712SW,128,116,-14,8,IAD,TPA,810,4,8,0,,0
1,3,4,926,1054,WN,1746,N612SW,88,78,-6,-4,IND,BWI,515,3,7,0,,0
1,3,4,1940,2121,WN,378,N726SW,101,87,11,25,IND,JAX,688,4,10,0,,0
1,3,4,1937,2037,WN,509,N763SW,240,230,57,67,IND,LAS,1591,3,7,0,,0
1,3,4,754,940,WN,1144,N778SW,226,205,-15,9,IND,PHX,1489,5,16,0,,0
1,3,4,1422,1657,WN,188,N215WN,155,143,47,87,ISP,FLL,1093,6,6,0,,0
1,3,4,1954,2239,WN,1754,N243WN,165,155,4,29,ISP,FLL,1093,3,7,0,,0
1,3,4,636,921,WN,2275,N454WN,165,147,-24,1,ISP,FLL,1093,5,13,0,,0
1,3,4,2107,2334,WN,362,N798SW,147,134,64,82,ISP,MCO,972,6,7,0,,0
1,3,4,1312,1546,WN,1397,N247WN,154,140,-4,12,ISP,MCO,972,7,7,0,,0
[zeppelin@sandbox data]#
```

Each column in the file can be interpreted using the guide below. The first comma-separated value in each line (index number 0) represents the month, the second value represents the day of the month, and so on. Of note for our purposes: the sixth value (index number 5) represents the carrier for each flight.

Lab 4: Create and Manipulate Pair RDDs (Scala)

Field	Index	Example data
Month	0	1
<u>DayofMonth</u>	1	3
<u>DayOfWeek</u>	2	4
<u>DepTime</u>	3	1738
<u>ArrTime</u>	4	1841
<u>UniqueCarrier</u>	5	WN
<u>FlightNum</u>	6	3948
<u>TailNum</u>	7	N467WN
<u>ElapsedTime</u>	8	63
<u>AirTime</u>	9	49
<u>ArrDelay</u>	10	1
<u>DepDelay</u>	11	8
Origin	12	JAX
<u>Dest</u>	13	FLL
Distance	14	318
<u>TaxiIn</u>	15	6
<u>TaxiOut</u>	16	8
Cancelled	17	0
<u>CancellationCode</u>	18	
Diverted	19	0

c. Use Zeppelin to import this file into the `/user/zeppelin` folder in HDFS.

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/flights.csv /user/zeppelin/flights.csv
```

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/flights.csv /user/zeppelin/flights.csv
```

**QUESTION:**

Why do this in Zeppelin instead of from the command line?

ANSWER:

When the tasks are performed in a Zeppelin notebook, the entire series of actions can be exported and then imported and replayed on another system. This will be discussed in more detail in a later lab exercise.

d. Create an RDD named `carrierRdd` by performing the following transformations:

1. Import the text file from HDFS using `sc.textFile()`.
2. Split the lines into an array of individual elements using `map()` (Hint: The elements are comma-separated rather than space-separated as in previous examples.)
3. Use `map()` to create a key-value pair from only the elements in the sixth column (index number 5) - which can be specified by appending `[5]` to the anonymous function value - and assign each instance a value of 1.
4. View the first five elements to confirm successful operation.

```
val carrierRdd = sc.textFile("/user/zeppelin/flights.csv").map(x =>
x.split(",")).map(column => (column(5),1))

carrierRdd.take(5)
```

```
val carrierRdd = sc.textFile("/user/zeppelin/flights.csv").map(x => x.split(",")).map(column => (column(5), 1))
carrierRdd.take(5)

carrierRdd: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[252] at map at <console>:29
res75: Array[(String, Int)] = Array((WN,1), (WN,1), (WN,1), (WN,1), (WN,1))
```

**NOTE:**

As in a previous example, these operations to create `carrierRdd` could have been performed in stages, using intermediate RDDs at each transformation step. We do not need the data in these intermediate forms, however, so chaining together multiple transformations to get to the final output works fine.

e. Perform a reduce and sort the results, then display the top three carrier codes by number of flights based on this data.

```
val carriersSorted = carrierRdd.reduceByKey((x,y) => x+y).map{case (a,b) =>
(b,a)}.sortByKey(ascending = false)

carriersSorted.take(3)
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

```
val carriersSorted = carrierRdd.reduceByKey((x,y) => x+y).map{case (a,b) => (b,a)}.sortByKey(ascending = false)
carriersSorted.take(3)

carriersSorted: org.apache.spark.rdd.RDD[(Int, String)] = ShuffledRDD[257] at sortByKey at <console>:31
res77: Array[(Int, String)] = Array((356167,WN), (175969,AA), (166445,00))
```

2. Determine the most common routes between two cities.

- The next exercise uses the `flights.csv` file from the previous lab, as well as the `airports.csv` file. Go back to the terminal window and take a look at the first few lines of the `airports.csv` file.

```
# head airports.csv
```

```
[zeppelin@sandbox data]# pwd
/home/zeppelin/spark/data
[zeppelin@sandbox data]# head airports.csv
iata,airport,city,state,country,lat,long
00M,Thigpen,BaySprings,MS,USA,31.95376472,-89.23450472
00R,LivingstonMunicipal,Livingston,TX,USA,30.68586111,-95.01792778
00V,MeadowLake,ColoradoSprings,CO,USA,38.94574889,-104.5698933
01G,Perry-Warsaw,Perry,NY,USA,42.74134667,-78.05208056
01J,HilliardAirpark,Hilliard,FL,USA,30.6880125,-81.90594389
01M,TishomingoCounty,Belmont,MS,USA,34.49166667,-88.20111111
02A,Gragg-Wade,Clanton,AL,USA,32.85048667,-86.61145333
02C,Capitol,Brookfield,WI,USA,43.08751,-88.17786917
02G,ColumbianaCounty,EastLiverpool,OH,USA,40.67331278,-80.64140639
[zeppelin@sandbox data]#
```

Each column in the file can be interpreted using the guide below. The first comma-separated value in each line (index number 0) represents the airport code, the second value represents the airport name, and so on. Of note for our purposes: the airport code (index number 0) and the airport city (index number 2).

Field	Index	Example
AirportCode	0	00M
Airport	1	Thigpen
City	2	Bay Springs
State	3	MS
Country	4	USA
Lat	5	31.95376472
Long	6	-89.23450472

Lab 4: Create and Manipulate Pair RDDs (Scala)

From the `flights.csv` file used earlier, columns 13 and 14 (index values 12 and 13) will be used in this exercise.

Field	Index	Example data
Origin	12	JAX
Dest	13	FLL

- b. Use Zeppelin to import the `airports.csv` file into the `/user/zeppelin` folder in HDFS.

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/airports.csv
/user/zeppelin/airports.csv
```

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/airports.csv /user/zeppelin/airports.csv
```

- c. Create an RDD named `cityRdd` by performing the following transformations:

1. Import the text file from HDFS using `sc.textFile()`.
2. Split the lines into an array of individual elements using `map()`
(Hint: Once again, the elements are comma-separated rather than space-separated.)
3. Use `map()` to pull out only the airport code and city elements in the first and third columns (index numbers 0 and 2).
4. View the first five elements to confirm successful operation.

```
val cityRdd = sc.textFile("/user/zeppelin/airports.csv").map(x =>
x.split(",")).map(column => (column(0), column(2)))

cityRdd.take(5)
```

```
val cityRdd = sc.textFile("/user/zeppelin/airports.csv").map(x => x.split(",")).map(column => (column(0), column(2))) FIN
cityRdd.take(5]

cityRdd: org.apache.spark.rdd.RDD[(String, String)] = MapPartitionsRDD[261] at map at <console>:29
res79: Array[(String, String)] = Array((iata,city), (00M,BaySprings), (00R,Livingston), (00V,ColoradoSprings), (01G,Perry))
```

d. Create an RDD named `flightOrigDestRdd` by performing the following transformations:

1. Import the text file from HDFS using `sc.textFile()`.
2. Split the lines into an array of individual elements using `map()`.
3. Use `map()` to pull out only the origin and destination elements in the 13th and 14th columns (index numbers 12 and 13).
4. View the first five elements to confirm successful operation.



NOTE:

Some of this code can be copied and pasted from a previous paragraph in the Zeppelin notebook.

```
val flightOrigDestRdd = sc.textFile("/user/zeppelin/flights.csv").map(x =>
x.split(",")).map(column => (column(12), column(13)))

flightOrigDestRdd.take(5)
```

```
val flightOrigDestRdd = sc.textFile("/user/zeppelin/flights.csv").map(x => x.split(",")).map(column => (column(12), column(13)))
flightOrigDestRdd.take(5)

FlightOrigDestRdd: org.apache.spark.rdd.RDD[(String, String)] = MapPartitionsRDD[265] at map at <console>:29
res81: Array[(String, String)] = Array((IAD,TPA), (IND,BWI), (IND,JAX), (IND,LAS), (IND,PHX))
```

e. Use `join()` to join `flightOrigDestRdd` and `cityRdd` into a third RDD named `origJoinRdd`.

This operation will result in an RDD that contains the origin code as the key, with a value of (destination code, origin city). This is half of the operation needed to get origin and destination cities.

View the first five elements to confirm successful operation.

```
val origJoinRdd = flightOrigDestRdd.join(cityRdd)

origJoinRdd.take(5)
```

```
val origJoinRdd = flightOrigDestRdd.join(cityRdd)
origJoinRdd.take(5)

origJoinRdd: org.apache.spark.rdd.RDD[(String, (String, String))] = MapPartitionsRDD[268] at join at <console>:33
res83: Array[(String, (String, String))] = Array((RIC,(CLE,Richmond)), (RIC,(EWR,Richmond)), (RIC,(EWR,Richmond)), (RIC,(EWR,Richmond)), (RIC,(EWR,Richmond)))
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

- f. Next use `join()` again to create an RDD named `destOrigJoinRdd` using `origJoinRdd` as a source and joining it `cityRdd` once again. Before performing the join operation, use `values()` to filter out the origin code (which is no longer needed) and pull out only the destination code and city name from the previous transformation.

This operation will result in an RDD that contains the destination code as the key, with a value of (origin city, destination city).

View the first five elements to confirm successful operation.

```
val destOrigJoinRdd = origJoinRdd.values.join(cityRdd)

destOrigJoinRdd.take(5)
```

```
val destOrigJoinRdd = origJoinRdd.values.join(cityRdd)
destOrigJoinRdd.take(5)

destOrigJoinRdd: org.apache.spark.rdd.RDD[(String, (String, String))] = MapPartitionsRDD[272] at join at <console>:35
res85: Array[(String, (String, String))] = Array((RIC,(Orlando,Richmond)), (RIC,(Orlando,Richmond)), (RIC,(Orlando,Ri
chmond)), (RIC,(Orlando,Richmond)))
```

- g. Create another RDD named `citiesCleanedRdd` that contains only the values of the `destOrigJoinRdd` (in other words, just the origin and destination city names). View the first five elements to confirm successful operation.

```
val citiesCleanedRdd = destOrigJoinRdd.values

citiesCleanedRdd.take(5)
```

```
val citiesCleanedRdd = destOrigJoinRdd.values
citiesCleanedRdd.take(5)

citiesCleanedRdd: org.apache.spark.rdd.RDD[(String, String)] = MapPartitionsRDD[273] at values at <console>:37
res87: Array[(String, String)] = Array((Philadelphia,Richmond), (Philadelphia,Richmond), (Philadelphia,Richmond), (Phi
ladelphia,Richmond))
```

- h. Use `map()` to convert the key-value pairs in `citiesCleanedRdd` into keys for a new RDD named `citiesKV`, and give each key a value of 1. View the first five elements to confirm successful operation.

```
val citiesKV = citiesCleanedRdd.map(cities => (cities, 1))

citiesKV.take(5)
```

```
val citiesKV = citiesCleanedRdd.map(cities => (cities, 1))
citiesKV.take(5)

citiesKV: org.apache.spark.rdd.RDD[((String, String), Int)] = MapPartitionsRDD[274] at map at <console>:39
res89: Array[((String, String), Int)] = Array(((Orlando,Richmond),1), ((Orlando,Richmond),1), ((Orlando,Richmond),1), (
Orlando,Richmond),1))
```

- i. Create an RDD named `citiesReducedSortedRdd` that reduces by key, swaps the keys and values, and then sorts by key in descending order. View the first three elements to confirm successful operation.

```
val citiesReducedSortedRdd = citiesKV.reduceByKey((x,y) => x+y).map{case (x,y) => (y,x)}.sortByKey(ascending = false)

citiesReducedSortedRdd.take(3)
```

```
val citiesReducedSortedRdd = citiesKV.reduceByKey((x,y) => x+y).map{case (x,y) => (y,x)}.sortByKey(ascending = false)
citiesReducedSortedRdd.take(3)

citiesReducedSortedRdd: org.apache.spark.rdd.RDD[(Int, (String, String))] = ShuffledRDD[279] at sortByKey at <console>:41
res91: Array[(Int, (String, String))] = Array((5540,(NewYork,Boston)), (5478,(Boston,NewYork)), (4183,(Chicago,NewYork)))
```



NOTE:

The top three origin city / destination combinations are New York to Boston, Boston to New York, and Chicago to New York.

3. Find the longest departure delays for any airline that experienced a delay of 15 minutes or more.

- a. This exercise once again uses the `flights.csv` file. This time we use the unique carrier code in column 6 (index value 5) and the departure delay value in minutes, which is in column 12 (index value 11).

Field	Index	Example data
UniqueCarrier	5	WN
DepDelay	11	8

- b. Create an RDD named `delayRdd` by performing the following transformations:

1. Import the `flights.csv` file from HDFS using `sc.textFile()`.
2. Split the lines into an array of individual elements using `map()`.
3. Use `filter()` to remove any lines for which the value of column 12 (index value 11) is less than 15. Because the `sc.textFile()` operation reads in all values as strings, you will need to cast the values in column 12 as integers prior to performing the `filter()` evaluation.
4. Use `map()` to pull out only the carrier code and departure delay elements in the 6th and 12th columns (index numbers 5 and 11). Store the values from column 12 as integers.

5. View the first five elements to confirm successful operation.

```
val delayRdd = sc.textFile("/user/zeppelin/flights.csv").map(x =>
x.split(",")).filter(delay => (delay(11).toInt) > 15).map(column => (column(5),
column(11).toInt))

delayRdd.take(5)
```

```
val delayRdd = sc.textFile("/user/zeppelin/flights.csv").map(x => x.split(",")).filter(delay => (delay(11).toInt) > 15).map(column => (column(5), column(11).toInt))
delayRdd.take(5)

delayRdd: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[289] at map at <console>:31
res96: Array[(String, Int)] = Array((MN,25), (WN,67), (MN,87), (MN,29), (MN,82))
```

For sake or readability, here is another screenshot of the above code with lines wrapped so that the code can be viewed in a larger font.

```
val delayRdd = sc.textFile("/user/zeppelin/flights.csv").map(x => x.split(",")).filter(delay => (delay(11).toInt) > 15
).map(column => (column(5), column(11).toInt))
delayRdd.take(5)

delayRdd: org.apache.spark.rdd.RDD[(String, Int)] = MapPartitionsRDD[289] at map at <console>:31
res96: Array[(String, Int)] = Array((MN,25), (WN,67), (MN,87), (MN,29), (MN,82))
```

- c. Create an RDD named `delayMaxRdd` that reduces the elements in `delayRdd` and returns only the longest delay per airline. For this exercise, it is not necessary to sort the values from largest to smallest.

Display five values to confirm successful operation.



REMINDER:

The reduce operation will need to compare all values for the same key and only keep the largest value in the final output.

The values in `delayRdd` were converted to integers in the previous step.

```
val delayMaxRdd = delayRdd.reduceByKey((x,y) => math.max(x, y))

delayMaxRdd.take(5)
```

```
val delayMaxRdd = delayRdd.reduceByKey((x,y) => math.max(x, y))
delayMaxRdd.take(5)

delayMaxRdd: org.apache.spark.rdd.RDD[(String, Int)] = ShuffledRDD[290] at reduceByKey at <console>:31
res98: Array[(String, Int)] = Array((B6,665), (FL,677), (OO,767), (AS,691), (UA,1268))
```


4. Remove records than contain incomplete data from a file.

- a. The next exercise uses the plane-data.csv. Go back to the terminal window and take a look at the first few lines of the plane-data.csv file.

```
# head plane-data.csv
```

```
[root@sandbox data]# pwd
/home/zeppelin/spark/data
[root@sandbox data]# head plane-data.csv
tailnum,type,manufacturer,issue_date,model,status,aircraft_type,engine_type,year
N050AA
N051AA
N052AA
N054AA
N055AA
N056AA
N057AA
N058AA
N059AA
[root@sandbox data]#
```

In the screenshot above, this file contains the column header names, followed by the column values. In this case, the first few records only have values for the first column, and the rest of the values are blank.

To see what complete records should look like, take a look at the last few lines of the file.

```
# tail plane-data.csv
```

```
[root@sandbox data]# tail plane-data.csv
N995AT,Corporation,BOEING,11/08/2002,717-200,Valid,Fixed Wing Multi-Engine,Turbo-Fan,2002
N995DL,Corporation,MCDONNELL DOUGLAS AIRCRAFT CO,03/06/1992,MD-88,Valid,Fixed Wing Multi-Engine,Turbo-Fan,1991
N996AT,Corporation,BOEING,07/30/2002,717-200,Valid,Fixed Wing Multi-Engine,Turbo-Fan,2002
N996DL,Corporation,MCDONNELL DOUGLAS AIRCRAFT CO,02/27/1992,MD-88,Valid,Fixed Wing Multi-Engine,Turbo-Fan,1991
N997AT,Corporation,BOEING,01/02/2003,717-200,Valid,Fixed Wing Multi-Engine,Turbo-Fan,2002
N997DL,Corporation,MCDONNELL DOUGLAS AIRCRAFT CO,03/11/1992,MD-88,Valid,Fixed Wing Multi-Engine,Turbo-Fan,1992
N998AT,Corporation,BOEING,01/23/2003,717-200,Valid,Fixed Wing Multi-Engine,Turbo-Fan,2002
N998DL,Corporation,MCDONNELL DOUGLAS CORPORATION,04/02/1992,MD-88,Valid,Fixed Wing Multi-Engine,Turbo-Jet,1992
N999CA,Foreign Corporation,CANADAIR,07/09/2008,CL-600-2B19,Valid,Fixed Wing Multi-Engine,Turbo-Jet,1998
N999DN,Corporation,MCDONNELL DOUGLAS CORPORATION,04/02/1992,MD-88,Valid,Fixed Wing Multi-Engine,Turbo-Jet,1992
[root@sandbox data]#
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

Each column in the file can be interpreted using the guide below. Note that there are nine possible column values for each record (index 0 through 8).

Field	Index	Example
<u>Tailnum</u>	0	N10156
Type	1	Corporation
Manufacturer	2	EMBRAER
<u>Issue_date</u>	3	02/13/2004
Model	4	EMB-145XR
Status	5	Valid
<u>Aircraft_type</u>	6	Fixed Wing Multi-Engine
<u>Engine_type</u>	7	Turbo-Fan
Year	8	2004

- b. Use Zeppelin to import the plane-data.csv file into the /user/zeppelin folder in HDFS.

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/plane-data.csv /user/zeppelin/plane-
data.csv
```

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/plane-data.csv /user/zeppelin/plane-data.csv
```

- c. Create an RDD named `planeDataRdd` from the `plane-data.csv` file. Before performing any transformations, use `count()` to display the number of lines in the RDD.

```
val planeDataRdd = sc.textFile("/user/zeppelin/plane-data.csv")
planeDataRdd.count()
```

```
val planeDataRdd = sc.textFile("/user/zeppelin/plane-data.csv")
planeDataRdd.count()

planeDataRdd: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[292] at textFile at <console>:29
res100: Long = 5030
```

Lab 4: Create and Manipulate Pair RDDs (Scala)

- d. Create an RDD named `cleanedPlaneDataRdd` by performing the following transformations:
1. Start with `planeDataRdd` from the previous step.
 2. Split the lines into an array of individual elements using `map()`. (Hint: The elements are comma-separated.)
 3. Use `filter()` to remove any lines that do not have a length of exactly 9 elements.
 4. Use `count()` to display the number of lines in the new RDD and confirm that the data set contains fewer lines than before.

```
val cleanedPlaneDataRdd = planeDataRdd.map(x => x.split(",")).filter(elements => elements.length == 9)

cleanedPlaneDataRdd.count()
```

```
val cleanedPlaneDataRdd = planeDataRdd.map(x => x.split(",")).filter(elements => elements.length == 9)
cleanedPlaneDataRdd.count()
```

```
cleanedPlaneDataRdd: org.apache.spark.rdd.RDD[Array[String]] = MapPartitionsRDD[294] at filter at <console>:31
```

Bonus Challenge Labs

The lab exercises below are for advanced students only. Instructor support and solutions will **not** be provided for these exercises. Some of the coding skills required to complete exercise 6 have not been covered in this class.

Perform the following steps:

1. **Extend CHALLENGE LABS exercise 4 by finding the top three most common airplane models for flights over 1500 miles.**

Both `flights.csv` and `plane-data.csv` will be used to solve this exercise.

2. **Extend CHALLENGE LABS exercises 1 and 3 by returning the names of the airlines rather than their carrier codes.**

To perform this extension, another file in the `/home/zeppelin/spark/data` directory must be used: `carriers.csv`. The data in this file contains two columns, as indicated below:

Field	Index	Example
Code	0	WN
Description	1	Southwest

This data contains additional challenges. The first row of the data contains column headers, just like `plane-data.csv` did. However, in addition, in some cases the description of the airline includes a comma that is not meant to separate values. For example, the airline with code `09Q` is has a description of `Swift Air, LLC`. The comma is part of the business name.

Good luck!

Lab 5: Basic Spark Streaming (Scala)

About This Lab

Objective:

Set up basic Spark Streaming operations using the REPL

File Locations:

/root/spark/data/

Successful Outcome:

Stream data from HDFS directories and TCP sockets using Spark Streaming

Lab Steps

Perform the following steps:

1. Use an HDFS directory as a streaming source.

- a. Open a terminal window and SSH into sandbox.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
```

- b. Create an HDFS directory for streaming output.

```
# hdfs dfs -mkdir /user/root/test/stream
```

- c. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- d. Set the log level to ERROR to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

Lab 5: Basic Spark Streaming (Scala)

- e. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

- f. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.  
g.StreamingContext@3e216fc9
```

- g. Create a DStream using `textFileStream()` to monitor the local HDFS directory `/user/root/test/`.

```
scala> val hdfsInputDS = sscFive.textFileStream("/user/root/test/")
```

```
scala> val hdfsInputDS = sscFive.textFileStream("/user/root/test/")  
hdfsInputDS: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spa  
rk.streaming.dstream.MappedDStream@4cf93f80
```

- h. Use `saveAsTextFiles()` to save the outputs to `/user/root/test/stream`.

```
scala> hdfsInputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> hdfsInputDS.saveAsTextFiles("/user/root/test/stream/")
```

- i. Print out the output to the terminal window.

```
scala> hdfsInputDS.print()
```

```
scala> hdfsInputDS.print()
```

- j. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```

Lab 5: Basic Spark Streaming (Scala)

```
scala> sscFive.start()

scala> -----
Time: 1464638880000 ms
-----

Time: 1464638885000 ms
-----
```

- k. Open a new terminal window, SSH to sandbox, and place the input file selfishgiant.txt from /root/spark/data into the folder. Observe what happens a few seconds later in the streaming terminal window.

```
# ssh sandbox
# hdfs dfs -put /root/spark/data/selfishgiant.txt /user/root/test/
```

```
root@ubuntu:~# ssh sandbox
```

```
[root@sandbox ~]# hdfs dfs -put /root/spark/data/selfishgiant.txt /user/root/test/selfishgiant.txt
```

```
sandbox:8020/user/root/.Trash/Current
root@sandbox ~]# hdfs dfs -put /root/spark/data/selfishgiant.txt /user/root/test/selfishgiant.txt
root@sandbox ~]#
```

```
Help
birds sat on the trees and sang
games in order to listen to th
ther.

k. He had been to visit his fr
seven years. After the seven y
, for his conversation was lim
e. When he arrived he saw the c
? he cried in a very gruff voice

?My own garden is my own garden,? said the Giant; ?any one
and I will allow nobody to play in it but myself.? So he
ound it, and put up a notice-board.

...

-----
Time: 1464639070000 ms
-----
```



NOTE:

You are free to upload additional files to see more streaming take place if you want.

Lab 5: Basic Spark Streaming (Scala)

- I. Once you observe data being streamed on-screen in the first terminal window, use the second terminal window to list the contents of the `/user/root/test/stream/` directory on HDFS.

```
#hdfs dfs -ls /user/root/test/stream/
```

```
[root@sandbox ~]# hdfs dfs -ls /user/root/test/stream/
```

-1464630920000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630925000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630930000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630935000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630940000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630945000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630950000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630955000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:55	/user/root/test/stream/name
-1464630960000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:56	/user/root/test/stream/name
-1464630965000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:56	/user/root/test/stream/name
-1464630970000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:56	/user/root/test/stream/name
-1464630975000	drwxr-xr-x	-	root	hdfs	0	2016-05-30	13:56	/user/root/test/stream/name

- m. In the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press enter.

```
sc.stop()  
exit()
```

2. Use a TCP socket as a streaming source.

- a. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- b. Set the log level to ERROR to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```


Lab 5: Basic Spark Streaming (Scala)

- c. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

- d. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.  
g.StreamingContext@3e216fc9
```

- e. Create a DStream using `socketTextStream()` to the system named “sandbox” on port 9999.

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)  
inputDS: org.apache.spark.streaming.dstream.ReceiverInputDStream[String] = org.a  
pache.spark.streaming.dstream.SocketInputDStream@4caeeab6
```

- f. Use `saveAsTextFiles()` to save the outputs to `/user/root/test/stream`.

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

- g. Print out the output to the terminal window.

```
scala> inputDS.print()
```

```
scala> inputDS.print()
```

- h. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```

Lab 5: Basic Spark Streaming (Scala)

```
at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
16/05/30 16:20:58 ERROR ReceiverTracker: Deregistered receiver for stream 0: Restarting receiver with delay 2000ms: Error connecting to sandbox:999 - java.net.ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
```



IMPORTANT:

An error will appear when the application starts because the application is waiting for an input connection.

- i. In the second terminal window use the netcat utility to create a connection to port 9999.

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

- j. Start typing words separated by space, hit Enter occasionally to submit them. Observe what happens in the streaming terminal window a few seconds after hitting Enter.

```
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
welcome to spark streaming

-----
Help
<init>(Socket.java:211)
c.streaming.dstream.SocketReceiver.receive(SocketInput
c.streaming.dstream.SocketReceiver$$anon$2.run(SocketI
-----
Time: 1464640255000 ns
-----
welcome to spark streaming
```

- k. Once you observe data being streamed on-screen in the first terminal window, use Ctrl + C (or Cmd + C if using a Mac) to exit netcat in the second terminal window.

```
[root@sandbox data]# nc -l kv 9999
Hello world
This is an example of streaming data
Random words random words
^C
[root@sandbox data]#
```

- l. Use the second terminal window to list the contents of the `/user/root/test/stream/` directory on HDFS. Note the time stamps on the files.

```
#hdfs dfs -ls /user/root/test/stream/
```

```
[root@sandbox ~]# hdfs dfs -ls /user/root/test/stream
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630920000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630925000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630930000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630935000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630940000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630945000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630950000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:55 /user/root/test/stream/name-1464630955000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:56 /user/root/test/stream/name-1464630960000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:56 /user/root/test/stream/name-1464630965000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:56 /user/root/test/stream/name-1464630970000
-rw-r--r-- 1 root hdfs 0 2016-05-30 13:56 /user/root/test/stream/name-1464630975000
```

- m. In the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press *Enter*.

```
sc.stop()
exit()
```

Result

You have created data streams from HDFS and TCP socket sources, observed the stream in real-time, and observed text files created from those streams for long-term storage and future use.

Lab 6: Basic Spark Streaming Transformations (Scala)

About This Lab

Objective:

Learn to use basic Spark Streaming transformations on data streams

File Locations:

/root/spark/data/

Successful Outcome:

Perform several basic transformations on streaming data

Lab Steps

Perform the following steps:

1. Perform a Spark Streaming transformations using flatmap().

- a. Open a terminal, connect to the sandbox cluster using SSH, and start a new instance of the REPL that is configured to use two CPU cores.

```
# ssh sandbox
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- b. Create a data stream the performs the following operations:

1. Sets the log level to "ERROR"
2. Imports the `StreamingContext` class
3. Creates an instance of that class named `sscFive` with a five-second time window
4. Creates a socket text `DStream` named `inputDS` that listens to "sandbox" on port 9999
5. Saves the `DStream` to text files in the `/user/root/test/stream/` directory.
6. Creates a `DStream` named `flatMapDS` that uses `flatMap()` to break lines into individual elements separated by spaces
7. Prints the contents of `flatMapDS` to the screen

8. Starts the application

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.StreamingContext@3e216fc9
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)  
inputDS: org.apache.spark.streaming.dstream.ReceiverInputDStream[String] = org.apache.spark.streaming.dstream.SocketInputDStream@1fb8b89e
```

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> val flatMapDS = inputDS.flatMap(_.split(" "))
```

```
scala> val flatMapDS = inputDS.flatMap(_.split(" "))  
flatMapDS: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.streaming.dstream.FlatMappedDStream@2cdb1581
```

```
scala> flatMapDS.print()
```

```
scala> flatMapDS.print()
```

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```

Lab 6: Basic Spark Streaming Transformations (Scala)

```
Starting receiver with delay 2000ms: Error connecting to sandbox:9999 - java.net.
ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)

-----
Time: 2016-05-31 06:57:40
-----
```



NOTE:

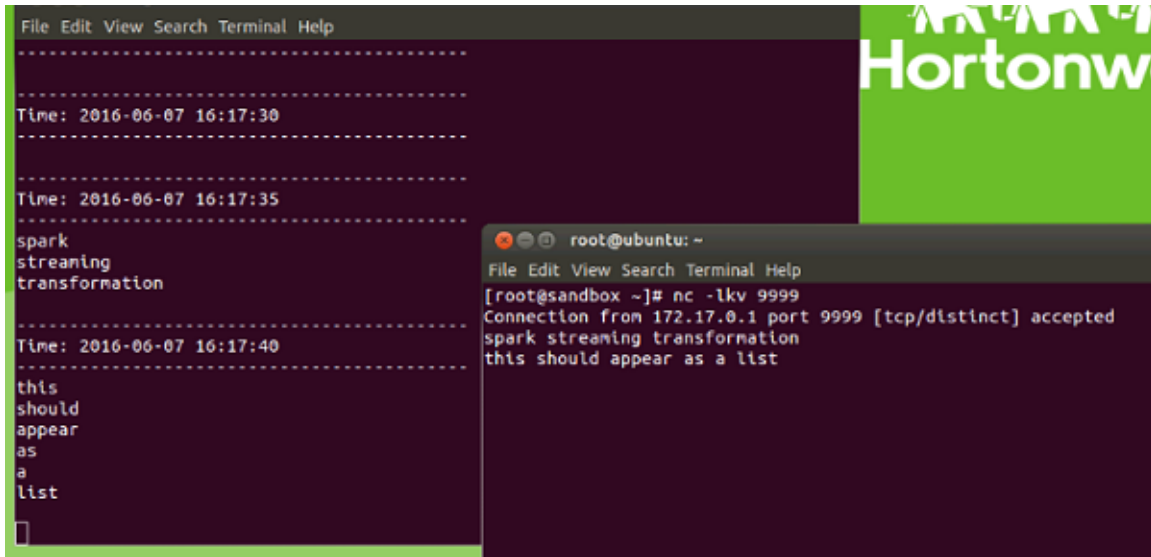
You will see an error when it starts because it is waiting for an input connection.

- c. Open a new terminal window, connect to the sandbox cluster, and connect to port 9999 using the netcat utility. Make sure both terminal windows are visible on-screen.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
```

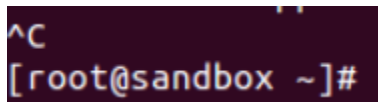
- d. In the netcat terminal, start typing words separated by spaces. Hit the Enter key occasionally to submit them to the stream. Observe how the words appear in the streaming window.



- e. In the streaming window, stop the stream and exit the REPL.

```
sc.stop()
exit()
```

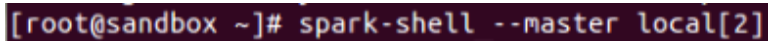
- f. In the netcat window, exit the socket by entering Ctrl + C (or CMD + C if using a Mac) on your keyboard.



2. Perform a Spark Streaming word count transformations using reduceByKey().

- a. In the streaming window, start a new instance of the REPL that is configured to use two CPU cores.

```
# spark-shell --master local[2]
```



- b. Create a data stream the performs the following operations:
 1. Sets the log level to "ERROR"
 2. Imports the StreamingContext class
 3. Creates an instance of that class named sscFive with a five-second time window
 4. Creates a socket text DStream named inputDS that listens to "sandbox" on port 9999

Lab 6: Basic Spark Streaming Transformations (Scala)

5. Saves the DStream to text files in the `/user/root/test/stream/` directory.
6. Creates a DStream named `wc` that uses `flatMap()`, `map()`, and `reduceByKey()` to count the number of times a word appears in a stream
7. Prints the contents of `wc` to the screen
8. Starts the application

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.  
g.StreamingContext@3e216fc9
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)  
inputDS: org.apache.spark.streaming.dstream.ReceiverInputDStream[String] = org.a  
pache.spark.streaming.dstream.SocketInputDStream@4caeeab6
```

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> inputDS.saveAsTextFiles("/user/root/test/stream/")
```

```
scala> val wc = inputDS.flatMap(_.split(" ")).map(word => (word,  
1)).reduceByKey((a,b) =>a+b)
```

```
scala> val wc = inputDS.flatMap(_.split(" ")).map(word => (word, 1)).reduceByKey  
((a,b) => a+b)  
wc: org.apache.spark.streaming.dstream.DStream[(String, Int)] = org.apache.spark  
.streaming.dstream.ShuffledDStream@33e31972
```

```
scala> wc.print()
```

```
scala> wc.print()
```

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```

```
Starting receiver with delay 2000ms: Error connecting to sandbox:9999 - java.net.
ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)

-----
Time: 2016-05-31 06:57:40
-----
```



NOTE:

You will see an error when it starts because it is waiting for an input connection.

- c. In the netcat window from the previous lab section, reconnect to port 9999 using the netcat utility. Make sure both terminal windows are visible on-screen.

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

In the netcat terminal, start typing words separated by spaces, making sure to repeat some of the words as you type. Hit the Enter key occasionally to submit them to the stream. Observe how the words appear in the streaming window.

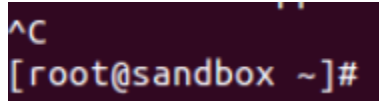
```
-----
Time: 2016-05-31 13:19:15
-----
Time: 2016-05-31 13:19:20
-----
Time: 2016-05-31 13:19:25
-----
(u'', 1)
(u'Spark', 1)
(u'to', 1)
(u'Welcome', 1)
(u'Streaming', 1)
-----
Time: 2016-05-31 13:19:30
-----
```

```
root@ubuntu: ~
File Edit View Search Terminal Help
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
Welcome to Spark Streaming
```

- e. In the streaming window, stop the stream and exit the REPL.

```
sc.stop()
exit()
```

- f. In the netcat window, exit the socket by entering Ctrl + C (or CMD + C if using a Mac) on your keyboard.

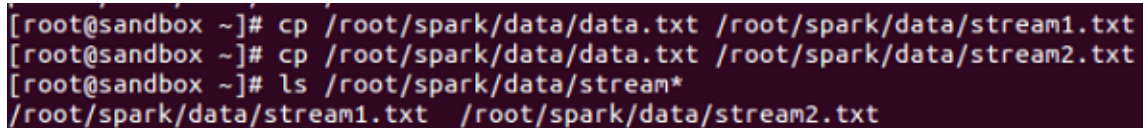


```
^C
[root@sandbox ~]#
```

3. Perform a Spark Streaming transformations using union().

- a. In the streaming window, create two copies of the `/root/spark/data/data.txt` file named `stream1.txt` and `stream2.txt` and confirm the operation was successful.

```
# cp /root/spark/data/data.txt /root/spark/data/stream1.txt
# cp /root/spark/data/data.txt /root/spark/data/stream2.txt
# ls /root/spark/data/stream*
```

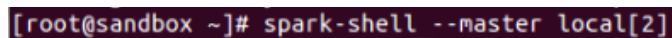


```
[root@sandbox ~]# cp /root/spark/data/data.txt /root/spark/data/stream1.txt
[root@sandbox ~]# cp /root/spark/data/data.txt /root/spark/data/stream2.txt
[root@sandbox ~]# ls /root/spark/data/stream*
/root/spark/data/stream1.txt /root/spark/data/stream2.txt
```

You can view the contents of the file if you want. As a reminder, these files contain a single line of text: "This is a test file"

- b. In the streaming window, start a new instance of the REPL that is once again configured to use two CPU cores.

```
# spark-shell --master local[2]
```



```
[root@sandbox ~]# spark-shell --master local[2]
```

- c. Create a data stream the performs the following operations:
1. Sets the log level to "ERROR"
 2. Imports the `StreamingContext` class
 3. Creates an instance of that class named `sscFive` with a five-second time window
 4. Creates two text file `DStream` named `inputDS1` and `inputDS2` that both listen to the `/user/root/test/` directory on HDFS.
 5. Creates a `DStream` named `combined` that uses `union()` to combine the two streams into a single `DStream`

6. Prints the contents of `combined` to the screen

7. Starts the application

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.  
g.StreamingContext@3e216fc9
```

```
scala> val inputDS1 = sscFive.textFileStream("/user/root/test/")
```

```
scala> val inputDS1 = sscFive.textFileStream("/user/root/test/")  
16/06/07 19:00:01 INFO FileInputDStream: Duration for remembering RDDs set to 60  
000 ms for org.apache.spark.streaming.dstream.FileInputDStream@70edffdb  
inputDS1: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.  
streaming.dstream.MappedDStream@7802fc4e
```

```
scala> val inputDS2 = sscFive.textFileStream("/user/root/test/")
```

```
scala> val inputDS2 = sscFive.textFileStream("/user/root/test/")  
16/06/07 19:00:53 INFO FileInputDStream: Duration for remembering RDDs set to 60  
000 ms for org.apache.spark.streaming.dstream.FileInputDStream@46e05384  
inputDS2: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.  
streaming.dstream.MappedDStream@27ae8ce3
```

```
scala> val combined = inputDS1.union(inputDS2)
```

```
scala> val combined = inputDS1.union(inputDS2)  
combined: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.  
streaming.dstream.UnionDStream@765d2d4d
```

```
scala> combined.print()
```

```
scala> combined.print()
```

```
scala> sscFive.start()
```

Lab 6: Basic Spark Streaming Transformations (Scala)

```
scala> sscFive.start()
```

- d. Go to the netcat terminal window (which we'll refer to now as the input1 window) from the previous lab section and type the command to upload the `small_blocks.txt` file from the local `/root/spark/data/` directory to the `/user/root/test/` directory on HDFS, but **DO NOT PRESS THE ENTER KEY**.

```
# hdfs dfs -put /root/spark/data/stream1.txt /user/root/test/
```

```
[root@sandbox ~]# hdfs dfs -put /root/spark/data/stream1.txt /user/root/test/
```

- e. Open a third terminal window (we'll refer to this as the input2 window), connect to the sandbox cluster, and type the same command as in the step above, but once again **DO NOT PRESS THE ENTER KEY**. Make sure both terminal windows are visible on-screen.

```
# ssh sandbox
```

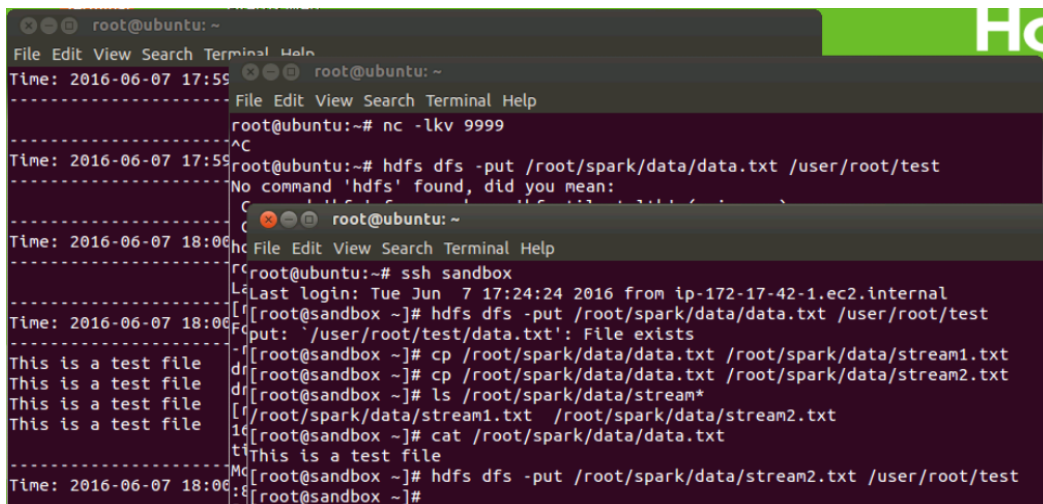
```
root@ubuntu:~# ssh sandbox
```

```
# hdfs dfs -put /root/spark/data/stream2.txt /user/root/test/
```

```
[root@sandbox ~]# hdfs dfs -put /root/spark/data/stream2.txt /user/root/test/
```

- f. Wait for a screen refresh in the streaming window, then immediately go to the input1 and input2 windows and press the Enter key.

Assuming you perform both actions within a 5-second collection window, the streaming window should display the contents of files as a combined data stream, as displayed in the screenshot below. The content of the text files (which in our case should be the same line of text) should each print multiple times because both streams were monitoring the same HDFS directory.



Lab 6: Basic Spark Streaming Transformations (Scala)

If your timing is off the first time, simply try again with a couple of additional copies that have unique file names like `streaming3.txt` and `streaming4.txt`.

- g. In the streaming window, stop the stream and exit the REPL.

```
sc.stop()
exit()
```

Result

You have successfully used several basic transformations on DStreams.

Lab 7: Spark Streaming Window Transformations (Scala)

About This Lab

Objective:

Use Spark Streaming Window Transformations

File Locations:

NA

Successful Outcome:

Perform several Spark Streaming Window Transformations

Lab Steps

Perform the following steps:

1. Create a streaming window using a TCP socket.
 - a. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- b. Set the log level to ERROR to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

- c. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._
import org.apache.spark.streaming._
```

- d. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.g.StreamingContext@3e216fc9
```

- e. Set the checkpoint directory.

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

- f. Create a DStream using `socketTextStream()` to the system named “sandbox” on port 9999 and set it up as a window function with a 15-second collection period (window length) and a 5-second collection interval.

```
scala> val inputDS =  
sscFive.socketTextStream("sandbox", 9999).window(Seconds(15), Seconds(5))
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999).window(Seconds(15),  
Seconds(5))  
inputDS: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.s  
treaming.dstream.WindowedDStream@6f53e863
```

- g. Print out the output to the terminal window.

```
scala> inputDS.print()
```

```
scala> inputDS.print()
```

- h. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```




NOTE:

An error will appear when the application starts because the application is waiting for an input connection.

```

        at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
16/05/30 16:20:58 ERROR ReceiverTracker: Deregistered receiver for stream 0: Restarting receiver with delay 2000ms: Error connecting to sandbox:999 - java.net.ConnectException: Connection refused
        at java.net.PlainSocketImpl.socketConnect(Native Method)
        at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
        at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
        at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
        at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
        at java.net.Socket.connect(Socket.java:589)
        at java.net.Socket.connect(Socket.java:538)
        at java.net.Socket.<init>(Socket.java:434)
        at java.net.Socket.<init>(Socket.java:211)
        at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
        at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
    
```

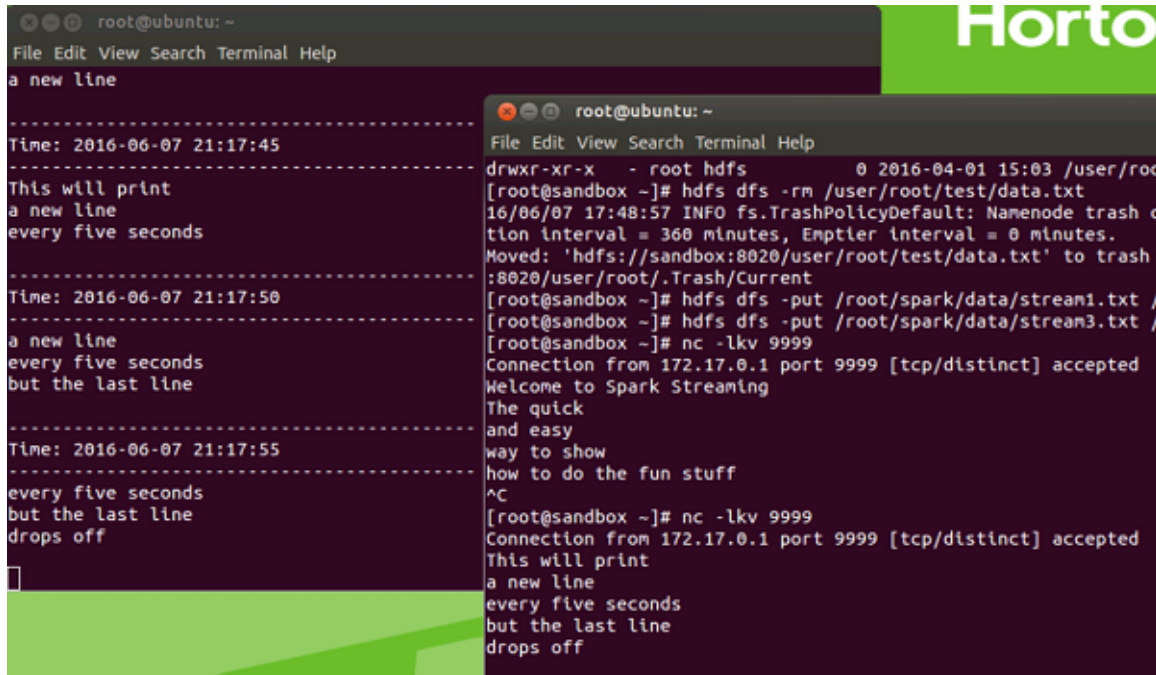
- i. In the second terminal window (connected to sandbox via SSH) use the netcat utility to create a connection to port 9999.

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

- j. Start typing words separated by space, hit **Enter** occasionally to submit them. Observe what happens in the streaming terminal window a few seconds after hitting **Enter**.

Lab 7: Spark Streaming Window Transformations (Scala)



```
root@ubuntu: ~
File Edit View Search Terminal Help
a new line
-----
Time: 2016-06-07 21:17:45
-----
This will print
a new line
every five seconds
-----
Time: 2016-06-07 21:17:50
-----
a new line
every five seconds
but the last line
-----
Time: 2016-06-07 21:17:55
-----
every five seconds
but the last line
drops off
[

root@ubuntu: ~
File Edit View Search Terminal Help
drwxr-xr-x  - root hdfs          0 2016-04-01 15:03 /user/root
[root@sandbox ~]# hdfs dfs -rm /user/root/test/data.txt
16/06/07 17:48:57 INFO fs.TrashPolicyDefault: Namenode trash c
tion interval = 360 minutes, Emptier interval = 0 minutes.
Moved: 'hdfs://sandbox:8020/user/root/test/data.txt' to trash
:8020/user/root/.Trash/Current
[root@sandbox ~]# hdfs dfs -put /root/spark/data/stream1.txt /
[root@sandbox ~]# hdfs dfs -put /root/spark/data/stream3.txt /
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
Welcome to Spark Streaming
The quick
and easy
way to show
how to do the fun stuff
^C
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
This will print
a new line
every five seconds
but the last line
drops off
```

- k. Once you observe data being streamed on-screen in the first terminal window, use **Ctrl + C** (or **Cmd + C** if using a Mac) to exit netcat in the second terminal window.

```
^C
[root@sandbox data]#
```

- l. In the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press **Enter**.

```
sc.stop()
exit()
```

2. Create a streaming window that counts words in a DStream using a TCP socket.

- a. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

Lab 7: Spark Streaming Window Transformations (Scala)

- b. Set the log level to `ERROR` to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

- c. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._
import org.apache.spark.streaming._
```

- d. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.StreamingContext@3e216fc9
```

- e. Set the checkpoint directory.

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

- f. Create a `DStream` using `socketTextStream()` to the system named "sandbox" on port 9999. Convert the lines of text it will accept into individual elements using `flatMap()`. Then use `countByWindow()` with a 15-second collection period (window length) and a 5-second collection interval to count the number of words typed over the last 15 seconds as a running total.

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999).flatMap(x => x.split(" "))
inputDS: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.streaming.dstream.MappedDStream@3a19b933
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999).flatMap(x => x.split(" "))
inputDS: org.apache.spark.streaming.dstream.DStream[String] = org.apache.spark.streaming.dstream.MappedDStream@3a19b933
```



QUESTION:

What do you think would happen if the `flatMap` function were removed from the line of code above?

- g. Print out the output to the terminal window.

```
scala> inputDS.print()
```

```
scala> inputDS.print()
```

- h. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```



NOTE:

An error will appear when the application starts because the application is waiting for an input connection.

```

    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketReceiver.scala:59)
16/05/30 16:20:58 ERROR ReceiverTracker: Deregistered receiver for stream 0: Restarting receiver with delay 2000ms: Error connecting to sandbox:999 - java.net.ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketReceiver.scala:59)

```

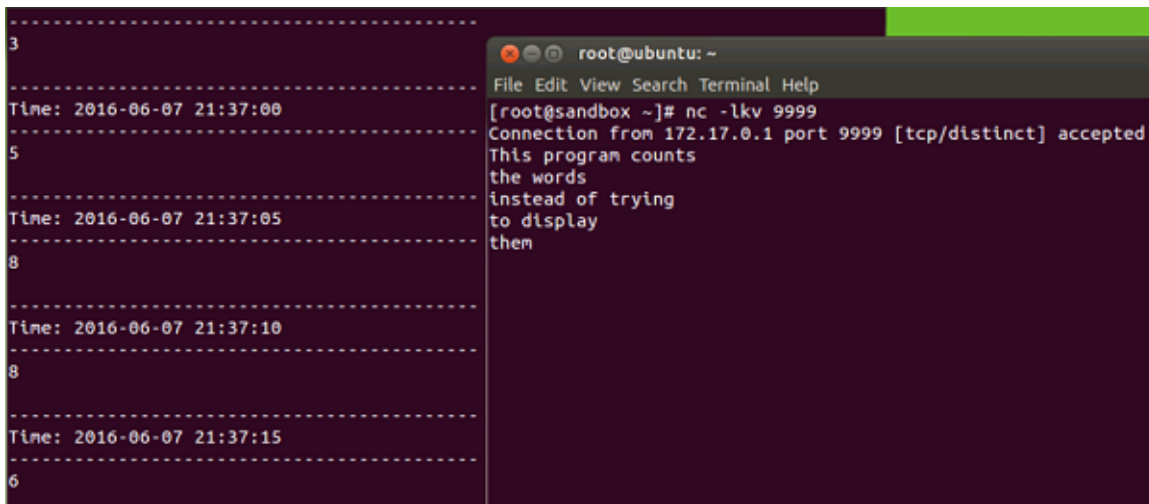
Lab 7: Spark Streaming Window Transformations (Scala)

- i. In the second terminal window use the netcat utility to create a connection to port 9999.

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

- j. Start typing words separated by space, hit **Enter** occasionally to submit them. Observe what happens in the streaming terminal window a few seconds after hitting **Enter**.



```
.....
3
.....
Time: 2016-06-07 21:37:00
.....
5
.....
Time: 2016-06-07 21:37:05
.....
8
.....
Time: 2016-06-07 21:37:10
.....
8
.....
Time: 2016-06-07 21:37:15
.....
6
.....

root@ubuntu: ~
File Edit View Search Terminal Help
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
This program counts
the words
instead of trying
to display
them
```

- k. Once you observe data being streamed on-screen in the first terminal window, use **Ctrl + C** (or **Cmd + C** if using a Mac) to exit netcat in the second terminal window.

```
^C
[root@sandbox data]#
```

- l. In the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press **Enter**.

```
sc.stop()
exit()
```

Lab 7: Spark Streaming Window Transformations (Scala)

3. Create a streaming window that counts instances of words in a DStream using a TCP socket.
 - a. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- b. Set the log level to ERROR to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

- c. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> import org.apache.spark.streaming._  
import org.apache.spark.streaming._
```

- d. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))  
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streamin  
g.StreamingContext@3e216fc9
```

- e. Set the checkpoint directory.

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

```
scala> sscFive.checkpoint("/user/root/test/checkpoint/")
```

Lab 7: Spark Streaming Window Transformations (Scala)

- f. Create a DStream using `socketTextStream()` to the system named “sandbox” on port 9999. Convert the lines of text it will accept into individual elements using `flatMap()`. Then use `map()` to create key-value pairs out of the individual elements. Finally, use `reduceByKeyAndWindow()` with a 15-second collection period (window length) and a 5-second collection interval to count the number of times a word has been typed over the last 15 seconds as a running total.

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999).flatMap(x =>
x.split(" ")).map(x => (x, 1)).reduceByKeyAndWindow((x,y) => x+y, (x,y) => x-y,
Seconds(15), Seconds(5))
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999).flatMap(x => x.sp
lit(" ")).map(x => (x, 1)).reduceByKeyAndWindow((x,y) => x+y, (x,y) => x-y, Seco
nds(15), Seconds(5))
inputDS: org.apache.spark.streaming.dstream.DStream[(String, Int)] = org.apache.
spark.streaming.dstream.ReducedWindowedDStream@93370c5
```

- g. Print out the output to the terminal window.

```
scala> inputDS.print()
```

```
scala> inputDS.print()
```

- h. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```



NOTE:

An error will appear when the application starts because the application is waiting for an input connection.

Lab 7: Spark Streaming Window Transformations (Scala)

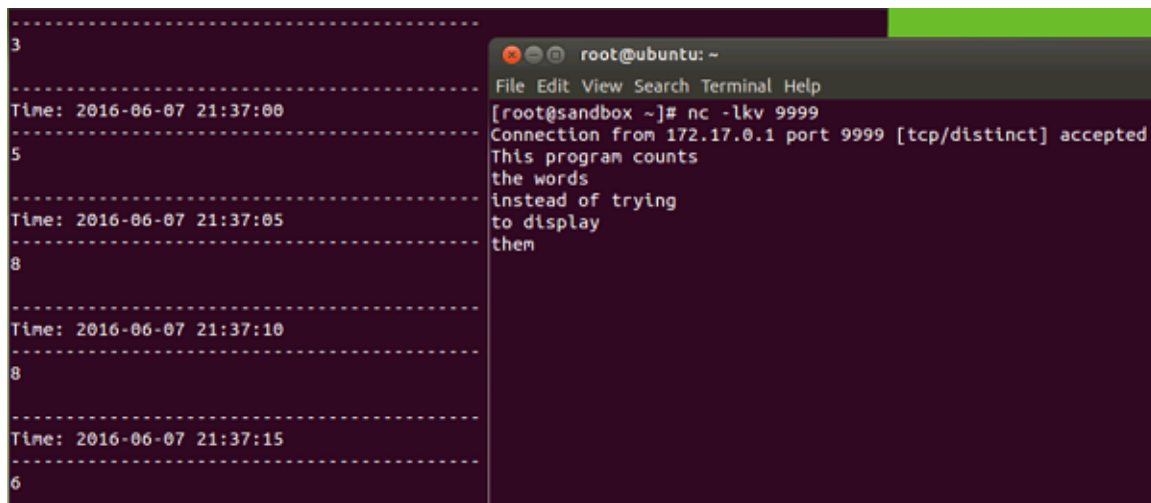
```
at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
16/05/30 16:20:58 ERROR ReceiverTracker: Deregistered receiver for stream 0: Restarting receiver with delay 2000ms: Error connecting to sandbox:999 - java.net.ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
```

- i. In the second terminal window use the netcat utility to create a connection to port 9999.

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

- j. Start typing words separated by space, hit **Enter** occasionally to submit them. Observe what happens in the streaming terminal window a few seconds after hitting **Enter**.

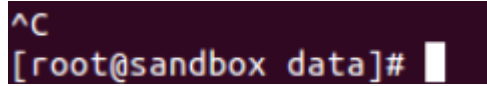


```
-----
3
-----
Time: 2016-06-07 21:37:00
-----
5
-----
Time: 2016-06-07 21:37:05
-----
8
-----
Time: 2016-06-07 21:37:10
-----
8
-----
Time: 2016-06-07 21:37:15
-----
6
```

```
root@ubuntu: ~
File Edit View Search Terminal Help
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
This program counts
the words
instead of trying
to display
them
```


Lab 7: Spark Streaming Window Transformations (Scala)

- k. Once you observe data being streamed on-screen in the first terminal window, use **Ctrl + C** (or **Cmd + C** if using a Mac) to exit netcat in the second terminal window.

A terminal window with a dark background. The first line shows a red caret (^) followed by a red C, representing the Ctrl+C command. The second line shows a shell prompt: [root@sandbox data]# followed by a white cursor bar.

- l. In the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press enter.

```
sc.stop()  
exit()
```

Result

You have successfully performed various spark Streaming Window Transformations

Lab 8: Create and Save DataFrames & Tables (Scala)

About This Lab

Objective:

Create and save DataFrames and tables

Files Locations:

NA

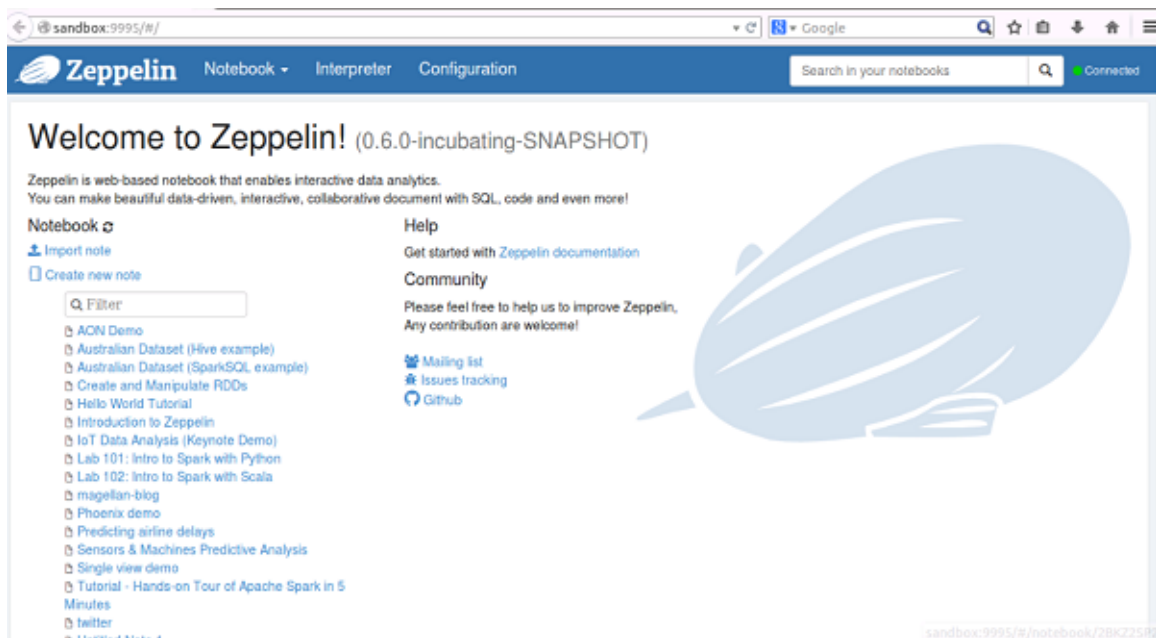
Successful Outcome:

Use various methods to create and save DataFrames and tables

Lab Steps

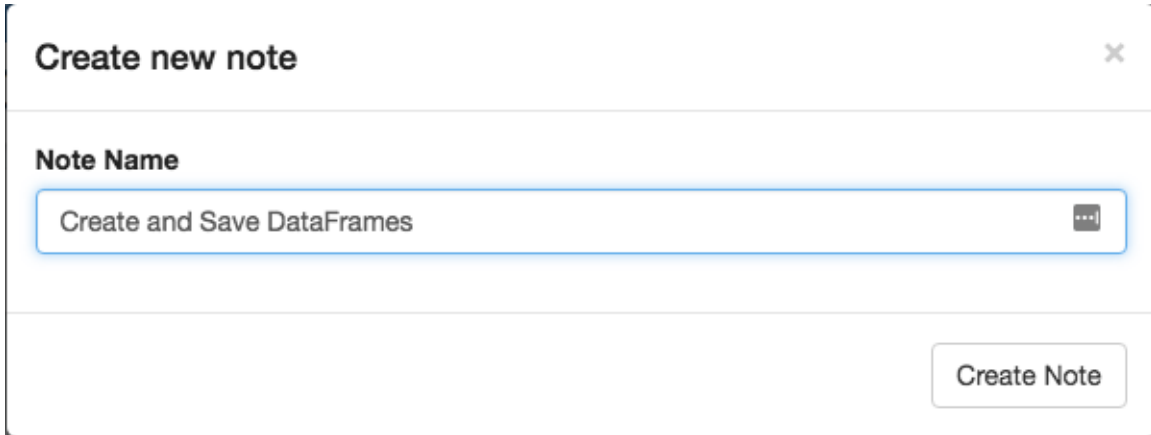
Perform the following steps:

1. Create and save DataFrames and tables.
 - a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.
<http://sandbox:9995/>



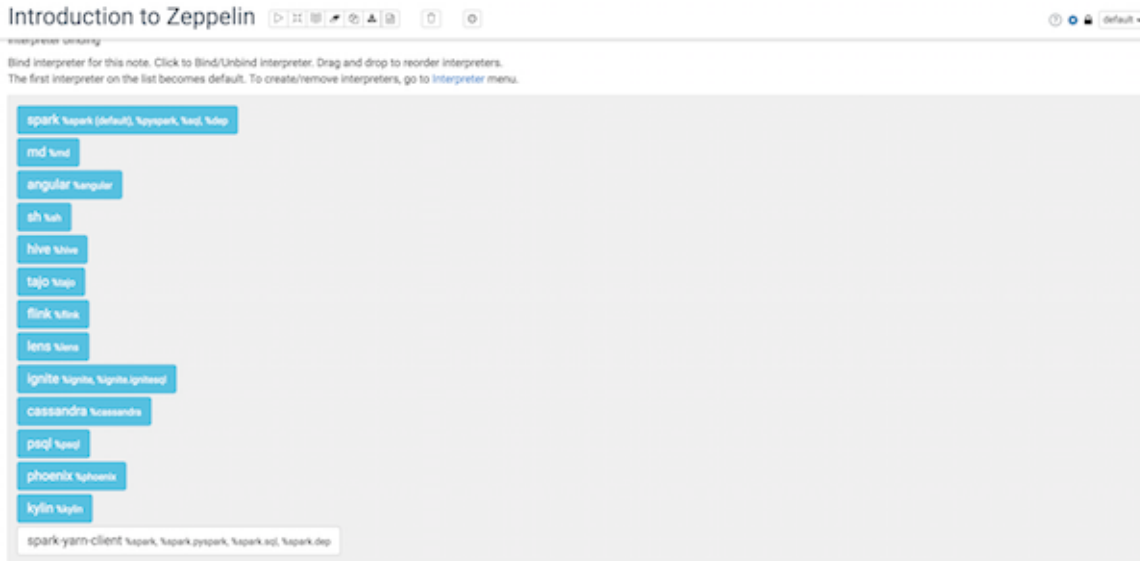
Lab 8: Create and Save DataFrames & Tables (Scala)

- b. Click Create new note. Name this note Create and Save DataFrames.



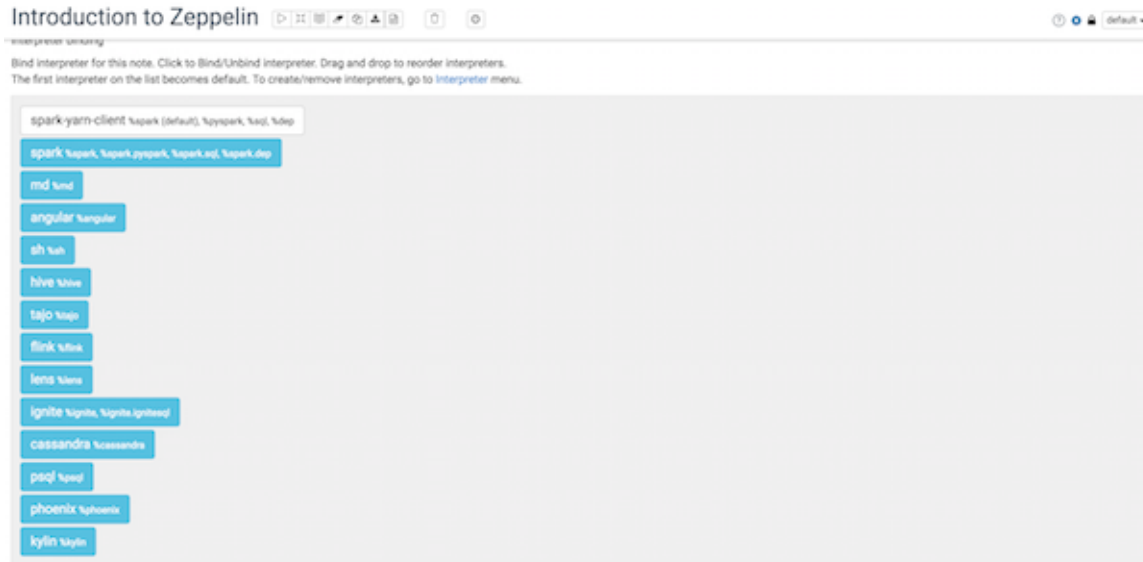
NOTE:
Make sure to set the interpreter to spark-yarn-client.

- c. At the top right click on the gear icon to change interpreter binding.

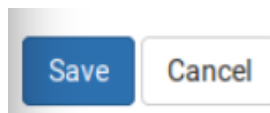


Lab 8: Create and Save DataFrames & Tables (Scala)

Drag the spark-yarn-client to the top and click save.



The first interpreter on the list becomes default.



d. Create a class named `DataSample` with the following attributes:

```
code: String, value: Long
```

Then create an RDD named `rddWithSchema` that utilizes this class to create a sequence of instances organized so that each element has a schema value.

The first entry in each instance should be a two-letter code (AA and BB). The second entry in each instance should be numeric values of 150,000 and 80,000 respectively that are assigned a schema value of `value`.

View the RDD to confirm success.

```
case class DataSample(code: String, value: Long)

val rddWithSchema = sc.parallelize(Seq(DataSample("AA", 150000),
DataSample("BB", 80000)))

rddWithSchema.collect()
```

Lab 8: Create and Save DataFrames & Tables (Scala)

```
case class DataSample(code: String, value: Long)
val rddWithSchema = sc.parallelize(Seq(DataSample("AA", 150000), DataSample("BB", 80000)))
rddWithSchema.collect()

defined class DataSample
rddWithSchema: org.apache.spark.rdd.RDD[DataSample] = ParallelCollectionRDD[783] at parallelize at <console>:92
res96: Array[DataSample] = Array(DataSample(AA,150000), DataSample(BB,80000))
```

- e. Use `toDF()` to convert this RDD to a new DataFrame named `dataframe2`. View the DataFrame to confirm success.

Note: The name `dataframe2` in this instruction is to keep the lab in sync with the Python version. In the Python version of the lab, an additional DataFrame named `dataframe1` is created prior to the equivalent of this step.

```
val dataframe2 = rddWithSchema.toDF()

dataframe2.show()
```

```
val dataframe2 = rddWithSchema.toDF()
dataframe2.show()
```

`dataframe2: org.apache.spark.sql.DataFrame = [code: string, value: bigint]`

code	value
AA	150000
BB	80000

- f. Register `dataframe2` as a temporary table named `table1temp`. Then issue a SQL command using the DataFrames API to show the tables visible to the context.

```
dataframe2.registerTempTable("table1temp")

sqlContext.sql("SHOW TABLES").show()
```

```
dataframe2.registerTempTable("table1temp")
sqlContext.sql("SHOW TABLES").show()
```

tableName	isTemporary
table1temp	true

- g. In the next paragraph, issue a Spark SQL command to `SHOW TABLES`. Does `table1temp` show up? If so, why? If not, why not?



NOTE:

Your output may also contain tables created if you ran demos in previous labs.

```
%sql
SHOW TABLES
```

%sql
SHOW TABLES

tableName	isTemporary
-----------	-------------

- h. Issue a HiveQL `CREATE TABLE` command from within the DataFrames API and create a permanent version of `table1temp` named `table1hive`. Use `SHOW TABLES` both from the DataFrames API, and then in a new paragraph from Spark SQL, to confirm this table is visible across contexts.

```
sqlContext.sql("CREATE TABLE table1hive AS SELECT * FROM table1temp")
sqlContext.sql("SHOW TABLES").show()
```


```
%sql
SHOW TABLES
```

```
sqlContext.sql("CREATE TABLE table1hive as SELECT * FROM table1temp")
sqlContext.sql("SHOW TABLES").show()

+-----+-----+
|  tableName|isTemporary|
+-----+-----+
|  table1temp|      true|
|  table1hive|     false|
+-----+-----+
```

Lab 8: Create and Save DataFrames & Tables (Scala)

```
%sql
SHOW TABLES
```




tableName	isTemporary
table1hive	false

- i. Use Spark SQL to view the contents of table1hive.

```
%sql
SELECT * FROM table1hive
```

```
%sql
SELECT * FROM table1hive
```



code	value
AA	150,000
BB	80,000

- j. Convert this Hive table into a DataFrame named dataframe3. View the new DataFrame to confirm success.

```
val dataframe3 = sqlContext.table("table1hive")
dataframe3.show()
```

```
val dataframe3 = sqlContext.table("table1hive")
dataframe3.show()

dataframe3: org.apache.spark.sql.DataFrame = [code: string, value: bigint]
+----+-----+
|code| value|
+----+-----+
| AA|150000|
| BB| 80000|
+----+-----+
```


- k. Save `dataframe3` to HDFS in JSON format to a folder named `dfJSON1`. In a new paragraph, list all contents of your HDFS home directory to confirm the DataFrame was successfully written.

```
dataframe3.write.format("json").save("dfJSON1")
```

```
%sh
```

```
hdfs dfs -ls dfJSON*
```

```
dataframe3.write.format("json").save("dfJSON1")
```

```
%sh
hdfs dfs -ls dfJSON*
```

Found 4 items

```
-rw-r--r--  3 zeppelin zeppelin      0 2016-06-12 14:24 dfJSON1/_SUCCESS
-rw-r--r--  3 zeppelin zeppelin    29 2016-06-12 14:24 dfJSON1/part-r-00000-96366c86-733e-49a3-b519-dbfbc21b13a7
-rw-r--r--  3 zeppelin zeppelin      0 2016-06-12 14:24 dfJSON1/part-r-00001-96366c86-733e-49a3-b519-dbfbc21b13a7
-rw-r--r--  3 zeppelin zeppelin    28 2016-06-12 14:24 dfJSON1/part-r-00002-96366c86-733e-49a3-b519-dbfbc21b13a7
```



NOTE:

The JSON file is stored in several `part-*` files in the folder name you specified. If you wanted to copy this file to your local file system for distribution outside the cluster, you could use `hdfs dfs -getmerge` to combine it as a single file on your local file system.

- l. View the combined contents of the files in the `dfJSON1` folder on HDFS.

```
%sh
```

```
hdfs dfs -cat dfJSON1/*
```

```
%sh
hdfs dfs -cat dfJSON1/*
```

```
{"code": "AA", "value": 150000}
{"code": "BB", "value": 80000}
```

**NOTE:**

The JSON format is not what you might typically see when looking at JSON files. For DataFrame creation, each row of information must be self-contained, and thus the formatting you see here is a requirement for converting JSON files to DataFrames. This same content coded in more typical JSON fashion would error out upon attempting to read it as a DataFrame.

- m. Create a new DataFrame named `dataframe4` from the contents of this folder on HDFS. View the new DataFrame to confirm success.

```
val dataframe4 = sqlContext.read.format("json").load("dfJSON1/*")
dataframe4.show()
```

```
val dataframe4 = sqlContext.read.format("json").load("dfJSON1/*")
dataframe4.show()

dataframe4: org.apache.spark.sql.DataFrame = [code: string, value: bigint]
+----+-----+
|code| value|
+----+-----+
| AA|150000|
| BB| 80000|
+----+-----+
```

Result

You have used several methods to create and save DataFrames and tables.

Lab 9: Working with DataFrames (Scala)

About This Lab

Objective:

Learn to use the DataFrames API.

File Locations:

NA

Successful Outcome:

Manipulate DataFrames using the DataFrames API

Lab Steps

Perform the following steps:

1. Manipulate DataFrames using the DataFrames API

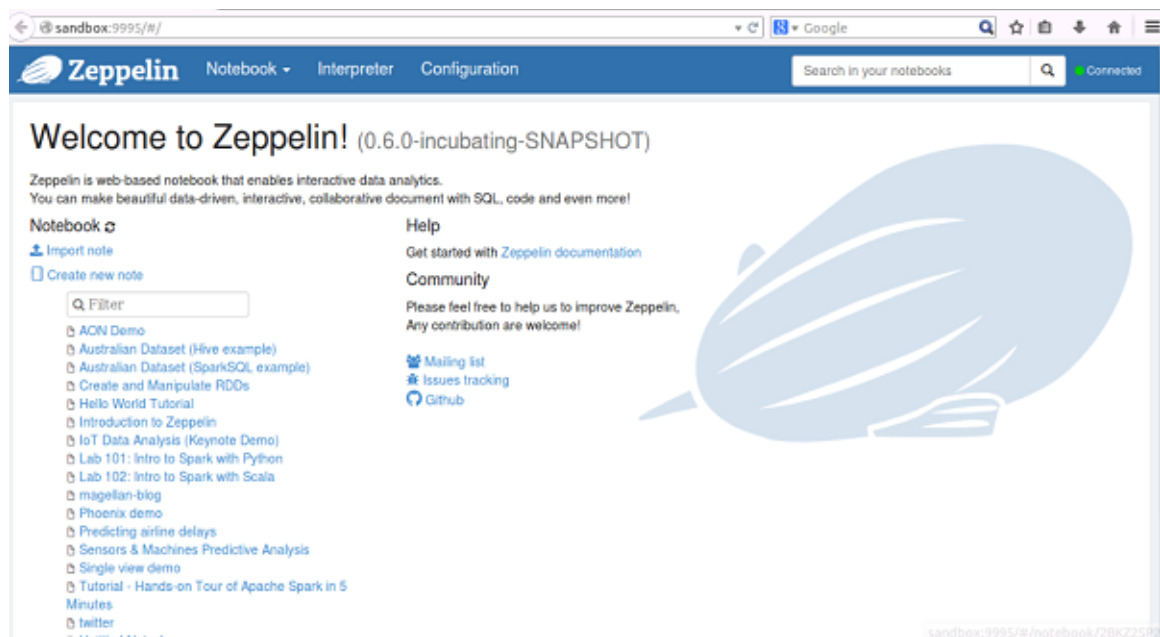


NOTE:

This lab intentionally makes use of one or more functions not discussed in the student book. The new functions are very similar in nature to functions already discussed in Core RDD programming and should make sense to the student. In addition, some functions are used in ways not discussed in the student book as well. This is to encourage exploration and experimentation, in addition to learning new ways to do things.

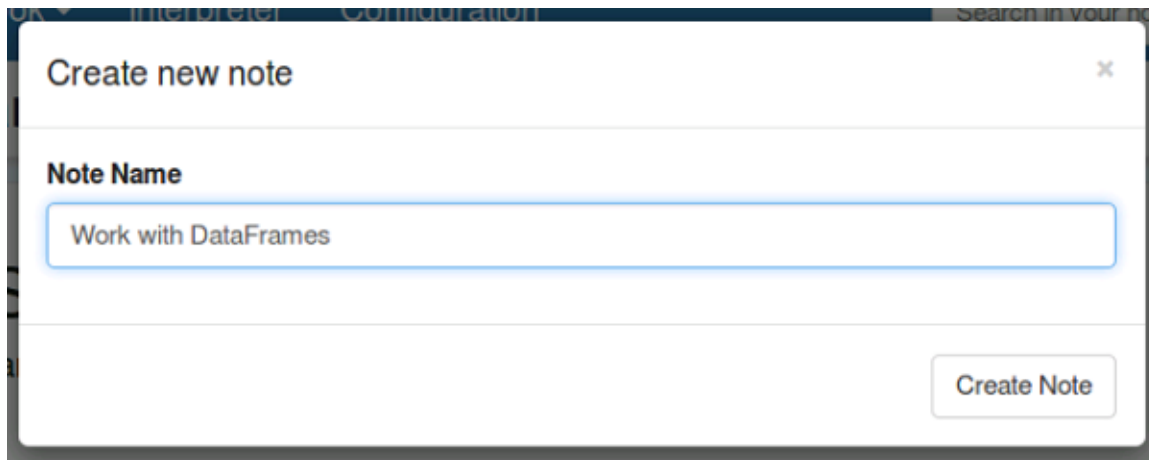
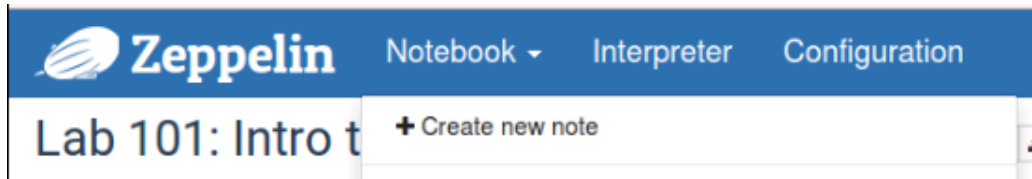
a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.

<http://sandbox:9995/>

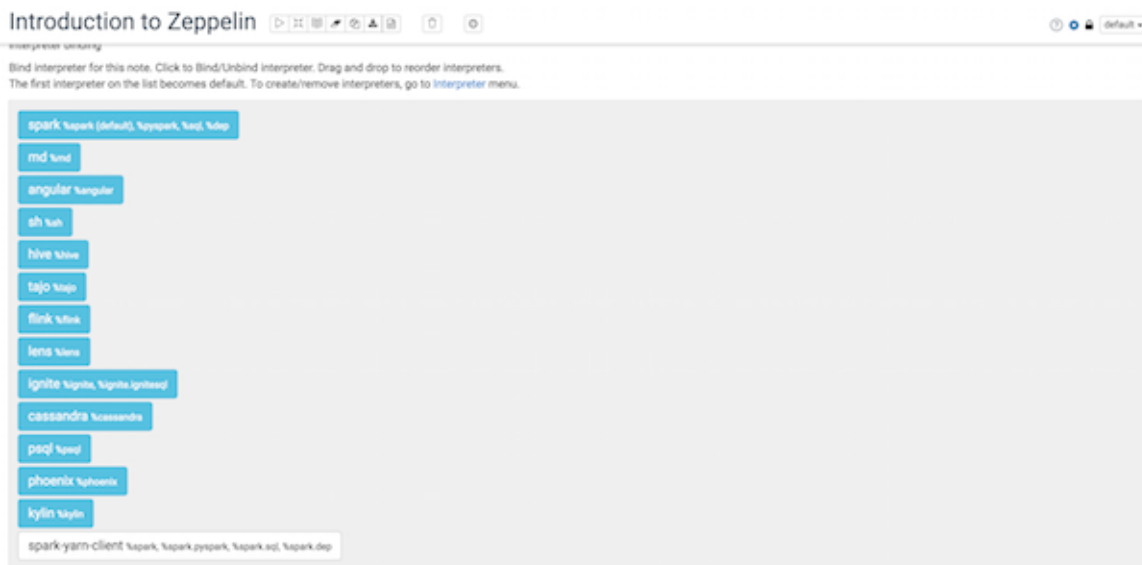


Lab 9: Working with DataFrames (Scala)

- b. Click on Notebook and select Create new note on the drop down. Name this note Work with DataFrames.

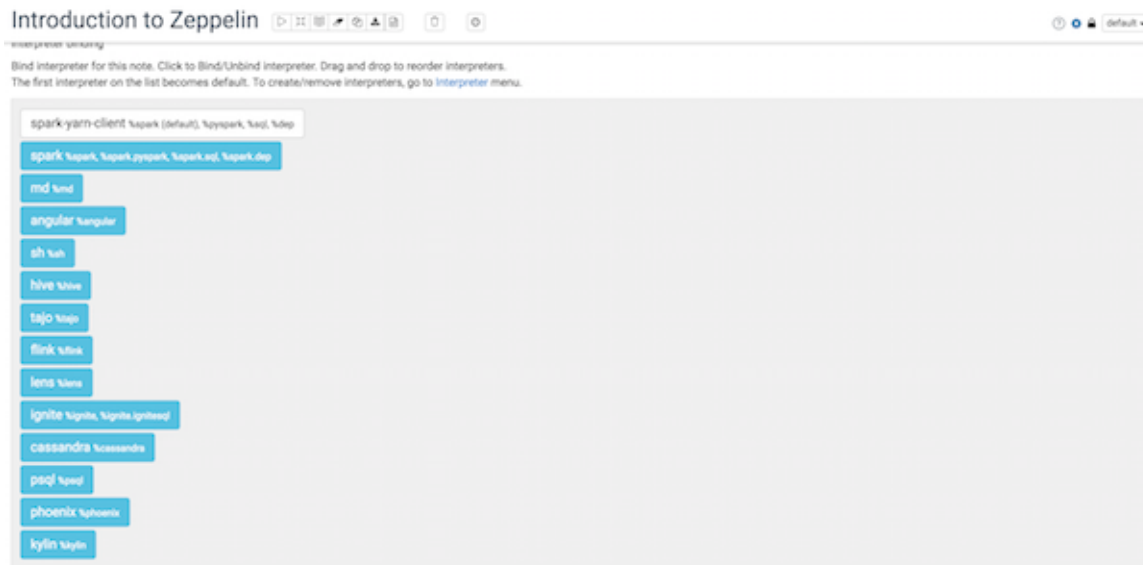


- c. At the top right click on the gear icon to change interpreter binding.

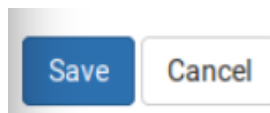


Lab 9: Working with DataFrames (Scala)

Drag the spark-yarn-client to the top and click save.



The first interpreter on the list becomes default.



- d. Create two DataFrames named `dataframeA` and `dataframeB` from the Hive table named `tablehive` created in the previous lab. Then use `unionAll()` to combine the rows of these two tables into a new DataFrame named `dataframeC`. Then show the contents of `dataframeC` to confirm success.

```
val dataframeA = sqlContext.table("tablehive")
val dataframeB = sqlContext.table("tablehive")
val dataframeC = dataframeA.unionAll(dataframeB)
dataframeC.show()
```

```
val dataframeA = sqlContext.table("table1hive")
val dataframeB = sqlContext.table("table1hive")
val dataframeC = dataframeA.unionAll(dataframeB)
dataframeC.show()
```

```
dataframeA: org.apache.spark.sql.DataFrame = [code: string, value: bigint]
dataframeB: org.apache.spark.sql.DataFrame = [code: string, value: bigint]
dataframeC: org.apache.spark.sql.DataFrame = [code: string, value: bigint]
+----+-----+
|code| value|
+----+-----+
| AA|150000|
| BB| 80000|
| AA|150000|
| BB| 80000|
+----+-----+
```

- e. Create a DataFrame named `dataframeD` that adds a column named `quarterly` that contains the contents of the `value` column multiplied by three. View the new DataFrame to confirm success.

```
val dataframeD = dataframeC.withColumn("quarterly", dataframeC("value") * 3)
dataframeD.show()
```

```
val dataframeD = dataframeC.withColumn("quarterly", dataframeC("value") * 3)
dataframeD.show()

dataframeD: org.apache.spark.sql.DataFrame = [code: string, value: bigint, quarterly: bigint]
+----+-----+-----+
|code| value|quarterly|
+----+-----+-----+
| AA|150000| 450000|
| BB| 80000| 240000|
| AA|150000| 450000|
| BB| 80000| 240000|
+----+-----+-----+
```

Lab 9: Working with DataFrames (Scala)

- f. Create a DataFrame named `dataframeE` that renames the `value` column to `monthly`. View the new DataFrame to confirm success.

```
val dataframeE = dataframeD.withColumnRenamed("value", "monthly")
dataframeE.show()
```

```
val dataframeE = dataframeD.withColumnRenamed("value", "monthly")
dataframeE.show()

dataframeE: org.apache.spark.sql.DataFrame = [code: string, monthly: bigint, quarterly: bigint]
+----+-----+-----+
|code|monthly|quarterly|
+----+-----+-----+
| AA| 150000| 450000|
| BB|  80000| 240000|
| AA| 150000| 450000|
| BB|  80000| 240000|
+----+-----+-----+
```

- g. Create a DataFrame named `dataframeF` that contains only those rows from `dataframeE` where the `quarterly` value is greater than 300,000. View the new DataFrame to confirm success.

```
val dataframeF = dataframeE.filter(dataframeE("quarterly") > 300000)
dataframeF.show()
```

```
val dataframeF = dataframeE.filter(dataframeE("quarterly") > 300000)
dataframeF.show()

dataframeF: org.apache.spark.sql.DataFrame = [code: string, monthly: bigint, quarterly: bigint]
+----+-----+-----+
|code|monthly|quarterly|
+----+-----+-----+
| AA| 150000| 450000|
| AA| 150000| 450000|
+----+-----+-----+
```

- h. Create a new DataFrame named `dataframeG` that adds the rows of `dataframeE` to `dataframeF` so that there are six rows total. View the new DataFrame to confirm success.

```
val dataframeG = dataframeE.unionAll(dataframeF)
dataframeG.show()
```

```
val dataframeG = dataframeE.unionAll(dataframeF)
dataframeG.show()

dataframeG: org.apache.spark.sql.DataFrame = [code: string, monthly: bigint, quarterly: bigint]
+-----+-----+-----+
|code|monthly|quarterly|
+-----+-----+-----+
| AA| 150000| 450000|
| BB|  80000| 240000|
| AA| 150000| 450000|
| BB|  80000| 240000|
| AA| 150000| 450000|
| AA| 150000| 450000|
+-----+-----+-----+
```

- i. Use `describe()` on `dataframeG` without supplying a column name and show the results.



QUESTION:

What happens?

```
dataframeG.describe().show()
```

```
dataframeG.describe().show()

+-----+-----+-----+
|summary|      monthly|      quarterly|
+-----+-----+-----+
|  count|              6|              6|
|   mean|126666.66666666667|      380000.0|
| stddev|36147.84456460256|108443.53369380768|
|   min|           80000|           240000|
|   max|          150000|          450000|
+-----+-----+-----+
```



ANSWER:

All columns with numeric values have statistics displayed.

- j. Show only unique rows from `DataFrameG`.

```
dataframeG.distinct().show()
```

```
dataframeG.distinct().show()
+----+-----+-----+
|code|monthly|quarterly|
+----+-----+-----+
| AA| 150000| 450000|
| BB| 80000| 240000|
+----+-----+-----+
```

- k. Use `drop()` to create a new `DataFrame` named `dataframeH` that contains only the `code` and `quarterly` columns. View the new `DataFrame` to confirm success.



QUESTION:

What other function described in the student book, could you have used to accomplish the same task? What would the code have been?

```
val dataframeH = dataframeG.drop("monthly")
dataframeH.show()
```

```
val dataframeH = dataframeG.drop("monthly")
dataframeH.show()

dataframeH: org.apache.spark.sql.DataFrame = [code: string, quarterly: bigint]
+----+-----+
|code|quarterly|
+----+-----+
| AA| 450000|
| BB| 240000|
| AA| 450000|
| BB| 240000|
| AA| 450000|
| AA| 450000|
+----+-----+
```

**ANSWER:**

The same thing could have been accomplished using the following code:

```
val dataframeH = dataframeG.select("code", "quarterly")
```

```
val dataframeH_Alt = dataframeG.select("code", "quarterly")
dataframeH_Alt.show()

dataframeH_Alt: org.apache.spark.sql.DataFrame = [code: string, quarterly: bigint]
+----+-----+
|code|quarterly|
+----+-----+
| AA|    450000|
| BB|    240000|
| AA|    450000|
| BB|    240000|
| AA|    450000|
| AA|    450000|
+----+-----+
```

- I. Create a new DataFrame named `dataframeI` that contains each unique element in the `code` column and a count of the number of times each code appears in `dataframeH`. View the new DataFrame to confirm success.

```
val dataframeI = dataframeH.groupBy("code").count()
dataframeI.show()
```

```
val dataframeI = dataframeH.groupBy("code").count()
dataframeI.show()

dataframeI: org.apache.spark.sql.DataFrame = [code: string, count: bigint]
+----+-----+
|code|count|
+----+-----+
| AA|     4|
| BB|     2|
+----+-----+
```

Result

You have successfully used the DataFrames API to manipulate DataFrames.

Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

About This Lab

Objective:

Learn to use Zeppelin to perform data visualizations, collaborate, and integrate visualizations into reports.

Files Locations:

NA

Successful Outcome:

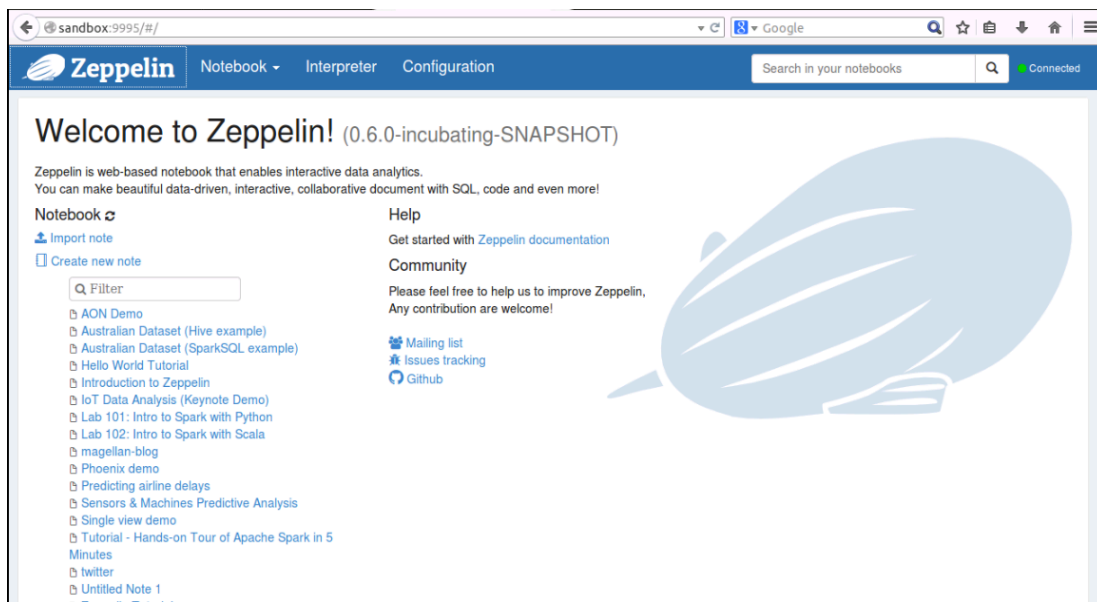
Use Zeppelin to perform data visualization, collaboration, and reporting tasks.

Lab Steps

Perform the following steps:

- 1 . Create data visualizations from a file of banking data.
 - a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.

<http://sandbox:9995/>

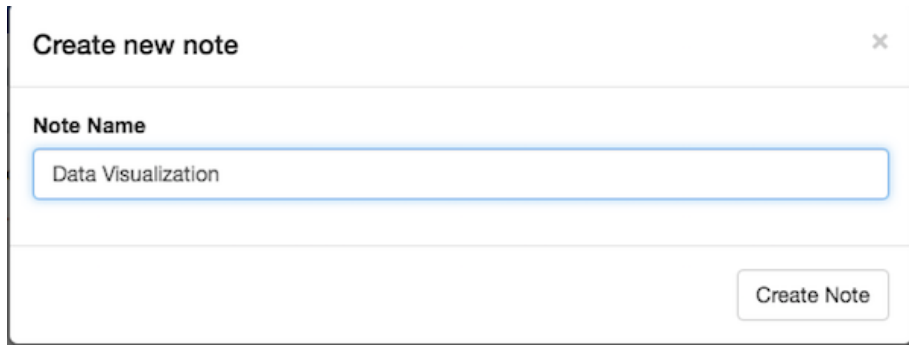


NOTE:

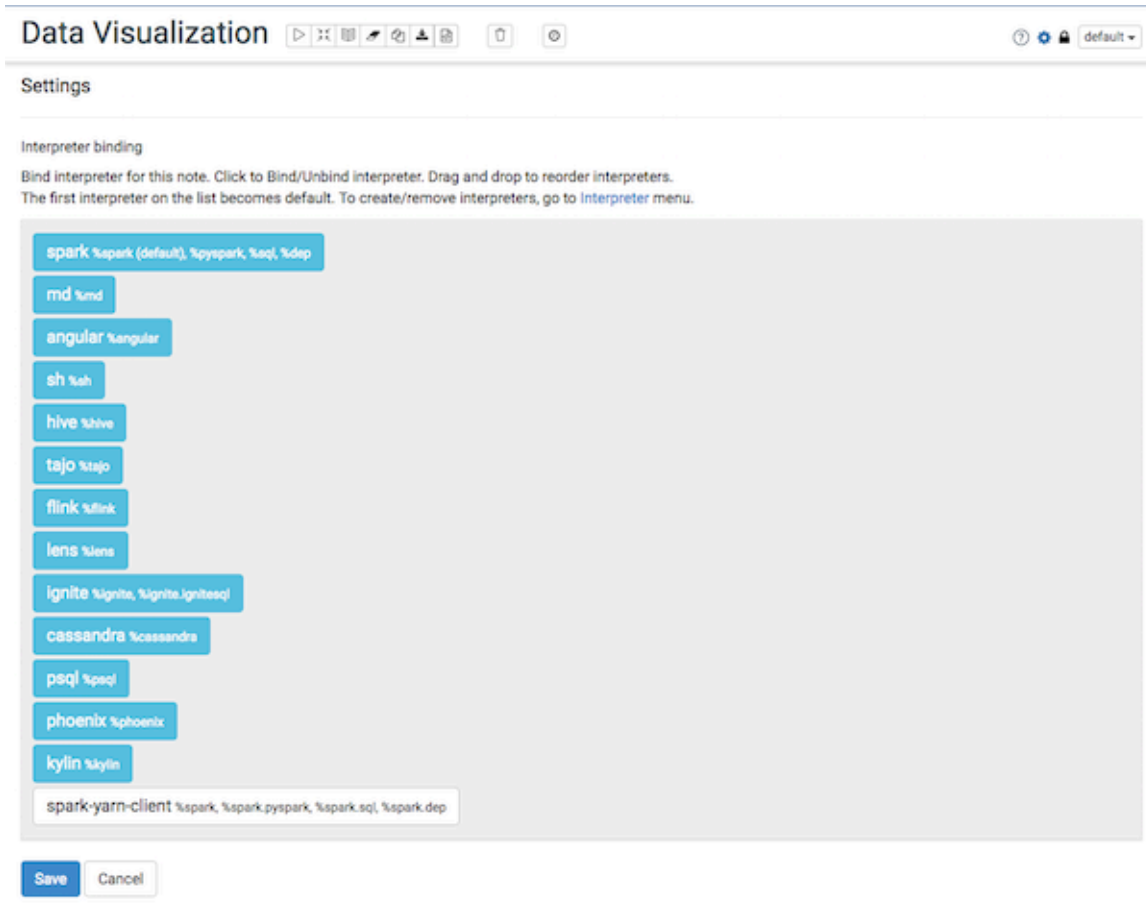
Zeppelin's current main backend processing engine is Apache Spark.

Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

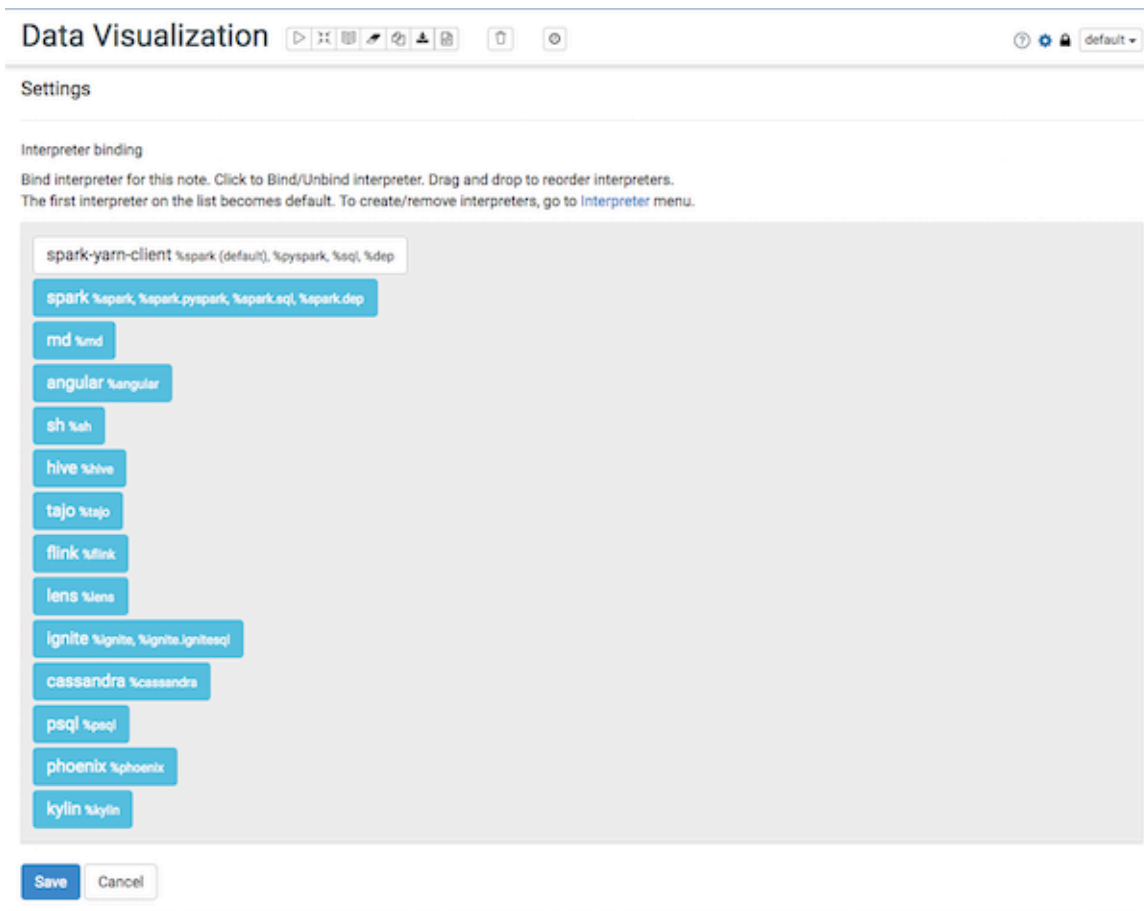
- b. Create a new note named Data Visualization.



- c. Set the interpreter for this note to spark-yarn-client.



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



The screenshot shows the Zeppelin Data Visualization interface. At the top, there is a header with the title "Data Visualization" and several icons. Below the header, there is a "Settings" section. Under "Settings", there is a sub-section for "Interpreter binding". The text below this sub-section reads: "Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu." Below this text is a list of interpreters, each represented by a blue button with its name and a placeholder for its configuration. The interpreters listed are: spark-yarn-client, spark, md, angular, sh, hive, tajo, flink, lens, ignite, cassandra, psq, phoenix, and kylin. At the bottom of the list, there are "Save" and "Cancel" buttons.

- d. Upload the bankdata3.orc file from the `/home/zeppelin/spark/data` directory on your local file system to your HDFS home directory. Confirm the file was uploaded successfully.

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/bankdata3.orc bankdata3.orc
hdfs dfs -ls bankdata*
```

```
%sh
hdfs dfs -put /home/zeppelin/spark/data/bankdata3.orc bankdata3.orc
hdfs dfs -ls bankdata*

-rw-r--r--  3 zeppelin zeppelin      1822 2016-06-06 12:07 bankdata3.orc
```



NOTE:

This data is a cleaned subset of a publicly available machine learning dataset. The original dataset can be found at the following link:

<http://archive.ics.uci.edu/ml/machine-learning-databases/00222/>

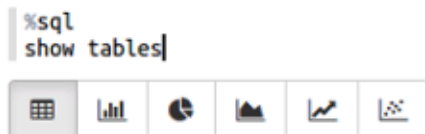
- e. Use the `bankdata3.orc` file to create a `DataFrame` named `bankdata`, a temporary table named `banktemp`, and a Hive table named `bankdataperm`.

```
val bankdata = sqlContext.read.format("orc").load("bankdata3.orc")
bankdata.registerTempTable("banktemp")
sqlContext.sql("create table bankdataperm as select * from banktemp")
```

```
val bankdata = sqlContext.read.format("orc").load("bankdata3.orc")
bankdata.registerTempTable("banktemp")
sqlContext.sql("create table bankdataperm as select * from banktemp")
```

- f. Use SQL to show the tables available and confirm that `bankdataperm` is available.

```
%sql
show tables
```



tableName	isTemporary
health_table	true
bankdataperm	false
tablehive	false

- g. Use SQL to select and display all rows and columns from `bankdataperm`.

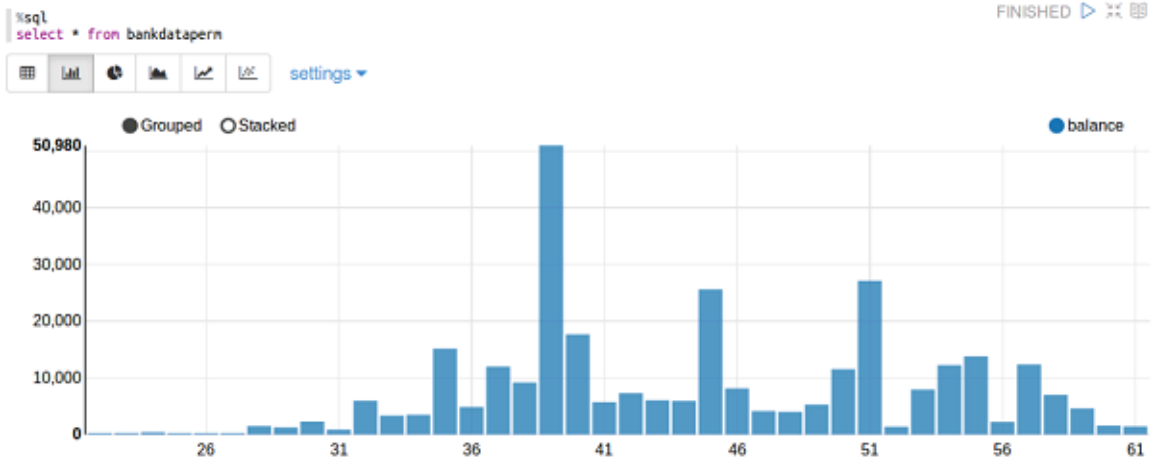
```
%sql
select * from bankdataperm
```

Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

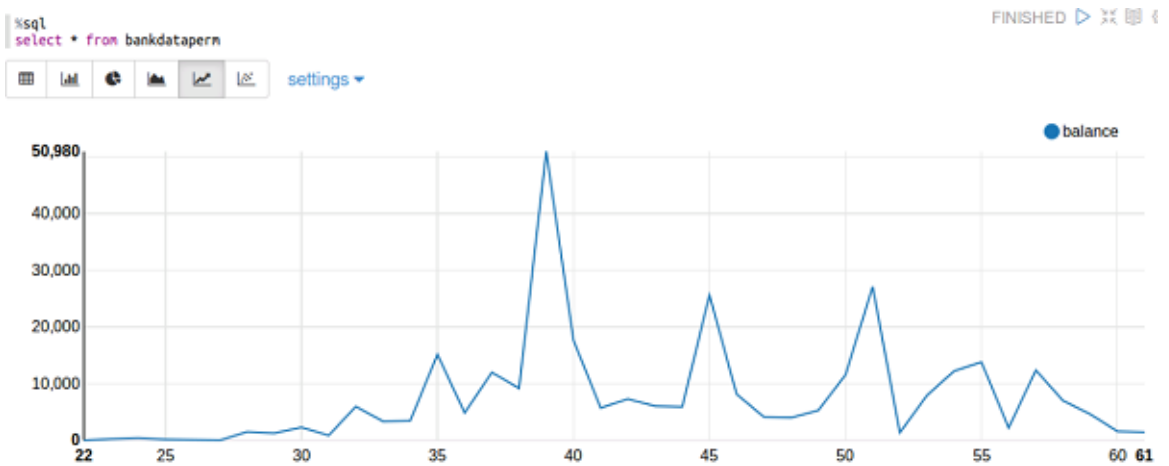
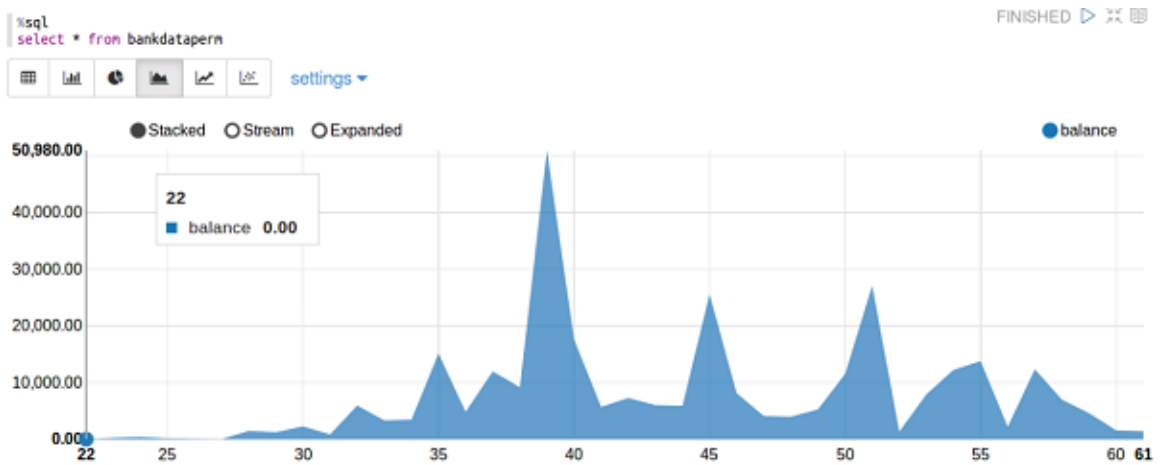
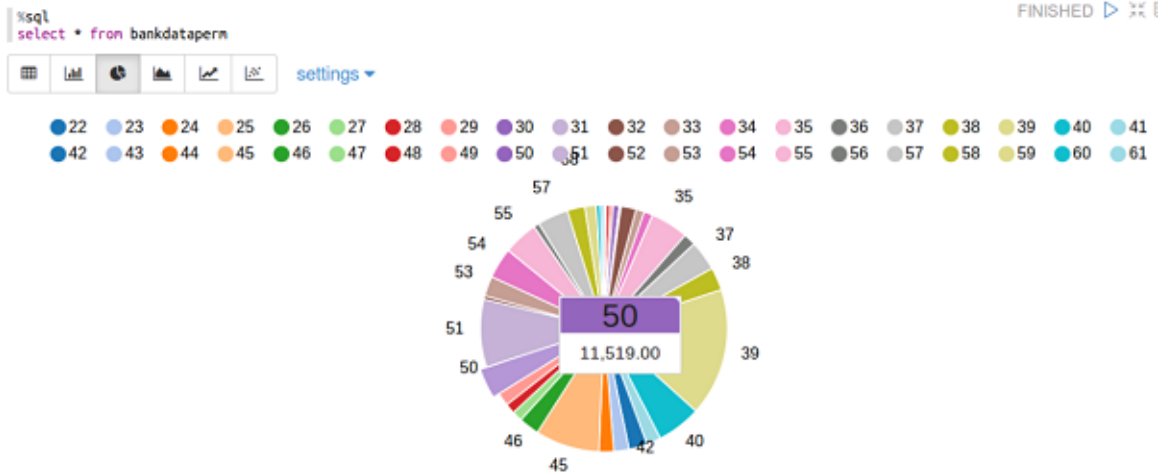
```
%sql
select * from bankdataperm
```

age	balance	marital
58	2,143	married
44	29	single
33	2	married
47	1,506	married
33	1	single

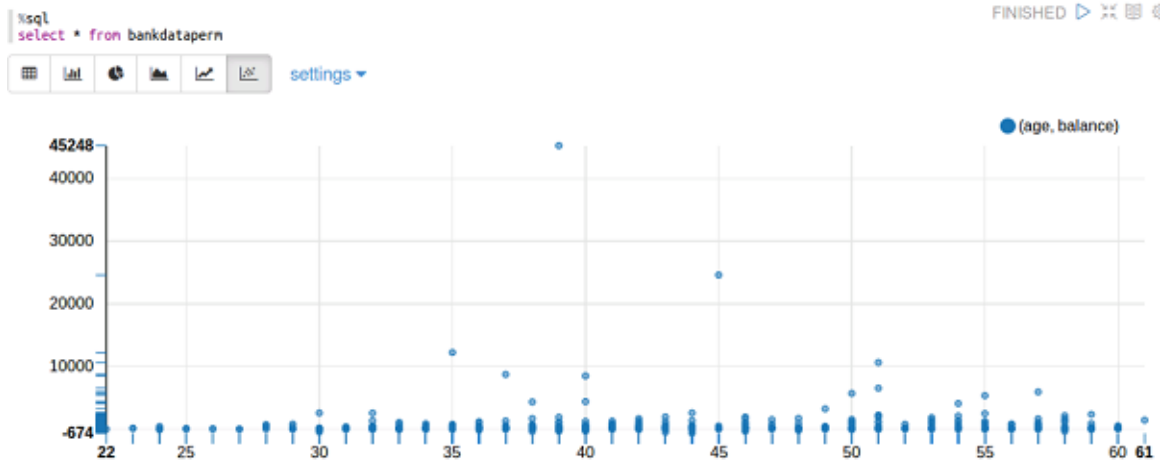
- h. Quickly browse through the five data visualizations available by default in Zeppelin. For most of this lab, we will work with the bar chart view.



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

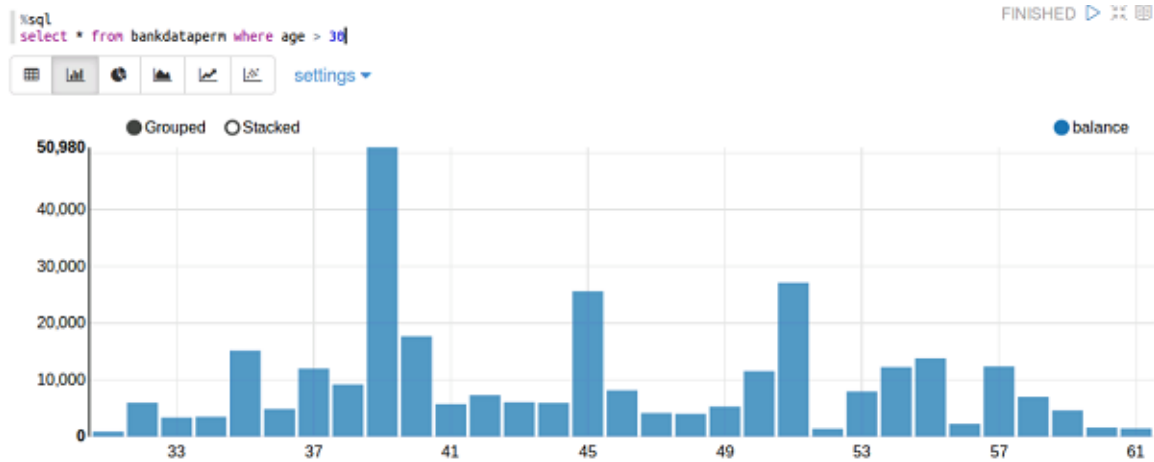


Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



- i. Go back to the bar chart view. Then, edit your SQL query so that it only shows data for individuals over the age of 30. Run the query and note the change in the chart.

```
%sql
select * from bankdatapern where age > 30
```



- j. Click on the settings link and notice that Zeppelin has selected the age column as the key column and is showing the sum of the balances for all individuals in each age bracket. Display the average balance instead of the sum of balances.

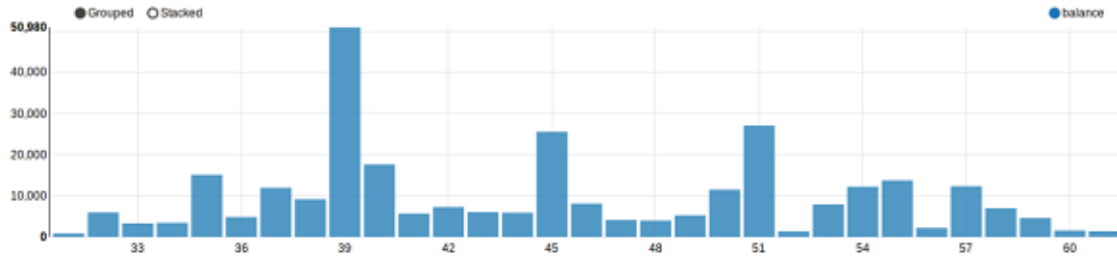
Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

All fields: age balance marital

Keys: age

Groups:

Values: balance SUM



Values

balance SUM

- sum
- count
- avg
- min
- max

Values

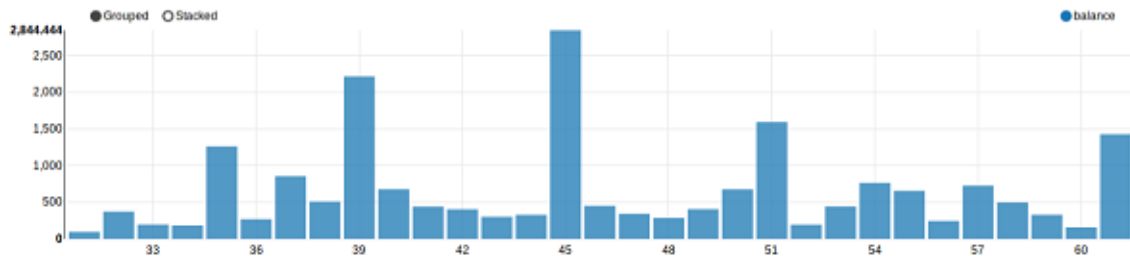
balance AVG

All fields: age balance marital

Keys: age

Groups:

Values: balance AVG



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

- k. Click and drag the available `marital` field into the `Groups` category to modify the visualization so that data is shown not only by age, but also grouped by marital status. When you are finished, click the settings link again to close the pivot chart options.

All fields: age balance marital

Keys: age

Groups:

Values: balance AVG

All fields: age balance marital

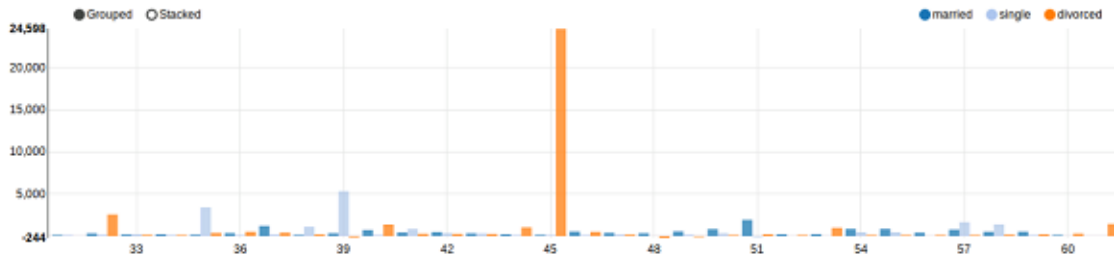
Groups: marital

All fields: age balance marital

Keys: age

Groups: marital

Values: balance AVG

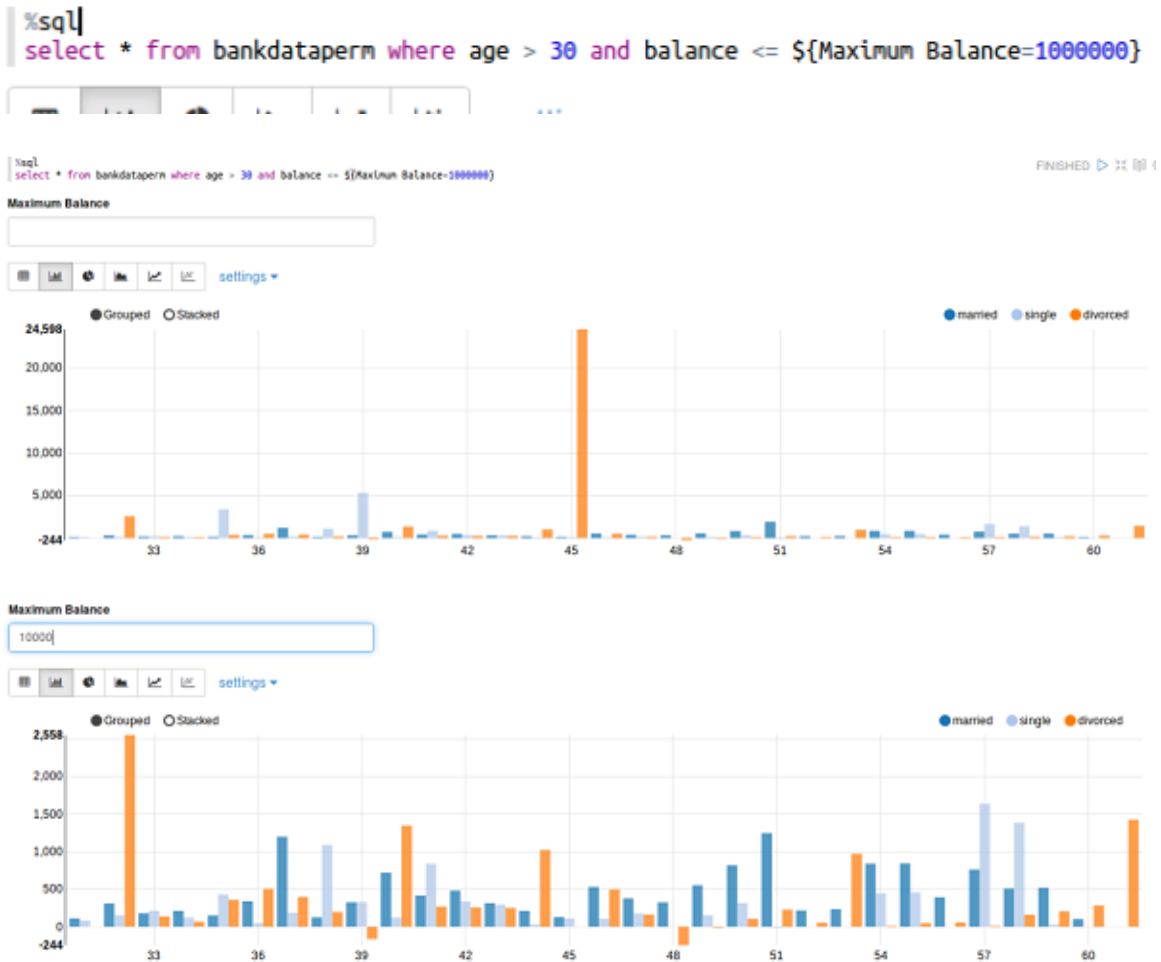


Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

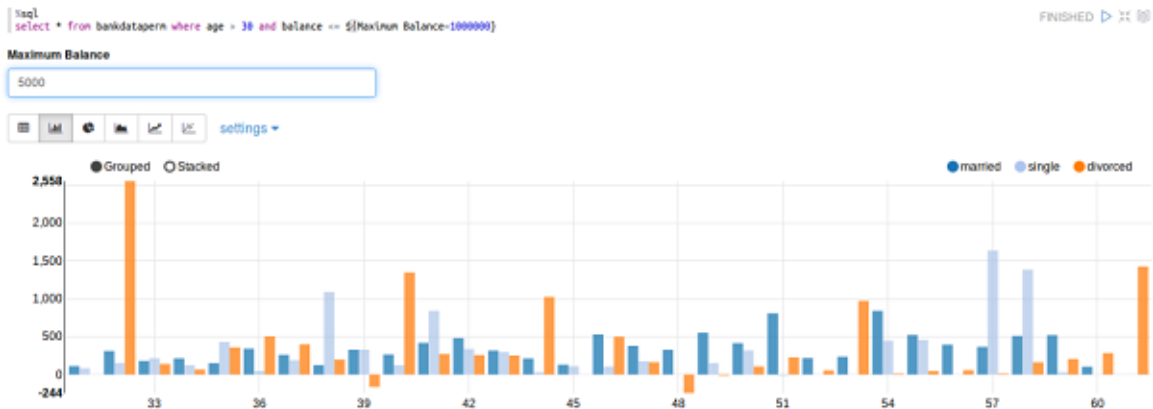
- I. It appears that we have what appears to be a single outlier that is skewing the data fairly significantly. We can easily see that the vast majority of average balances are well below \$5,000. Add a dynamic form to the SQL query that allows you to filter out data where the maximum balance for any individual exceeds a certain threshold, but set the default to 1,000,000 so that it doesn't immediately modify the chart. Rerun the query with this new code, then use this dynamic form to adjust the maximum balance to \$10,000 and \$5,000 and note the effects on the visualization.

```
%sql
```

```
select * from bankdataperm where age > 30 and balance <= ${Maximum  
Balance=1000000}
```



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

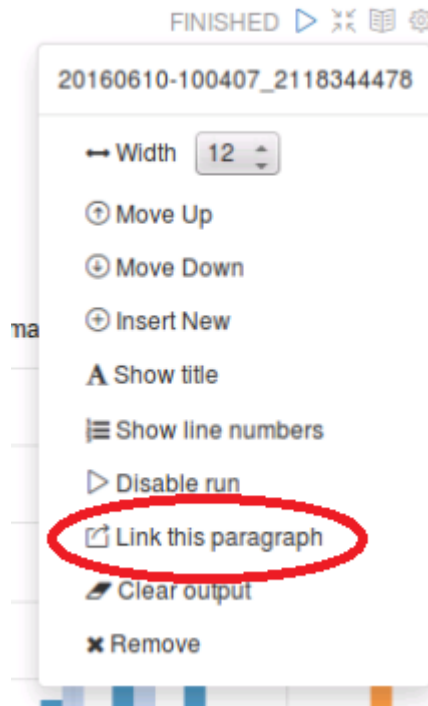


QUESTIONS:

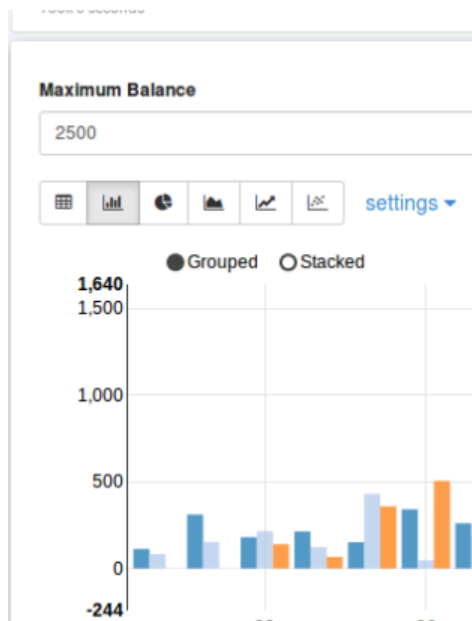
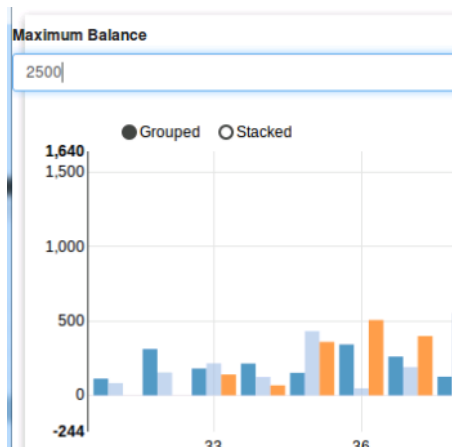
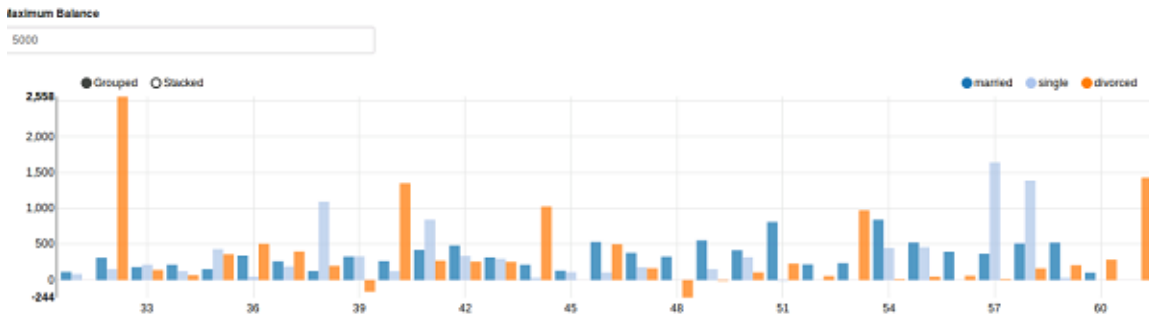
Why do you think changing the maximum balance from \$10,000 to \$5,000 had so little effect on the chart?

What group (married, single, or divorced) had the most change based on changing the maximum balance?

- m. Create a URL that allows you to share this chart with others without giving them access to the code or the Zeppelin note. Use the linked page to change the maximum balance to \$2,500, then return to your note and observe the effects the change had at the source.



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)

- n. In the paragraph below this one, run the SQL command to read all data from `bankdataperm`. Then adjust the width of the two paragraphs so that they both appear on the same line.



Took 0 seconds (outdated)

```
%sql  
select * from bankdataperm
```



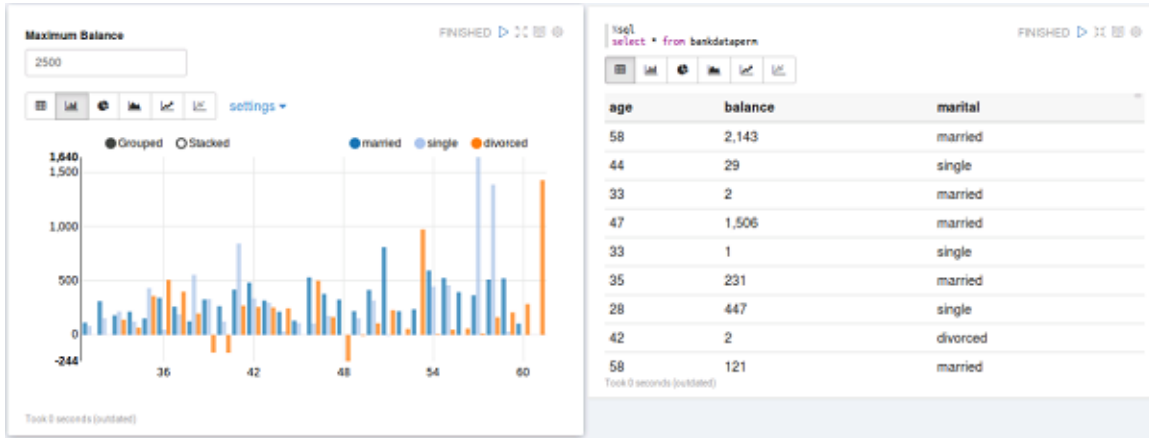
age	balance	marital
58	2,143	married
44	29	single
33	2	married
47	1,506	married

FINISHED ▶ ⌘ 📖 ⚙

20160610-101920_468564635

- ← Width 12
- ⊕ Move U 1
- ⊕ Move D 2
- ⊕ Insert N 3
- ⊕ Insert N 4
- ⊕ Insert N 5
- ⊕ Show ti 6
- ☰ Show li 7
- ▶ Disable 8
- 🔗 Link th 9
- 🗑 Clear output 10

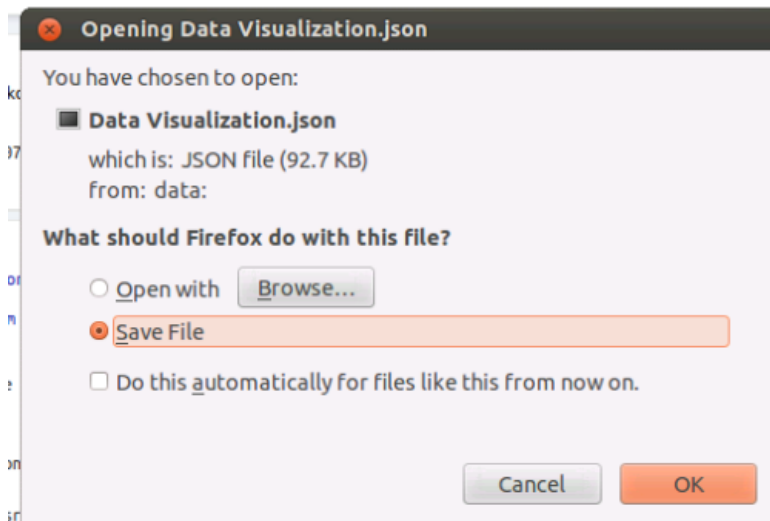
Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



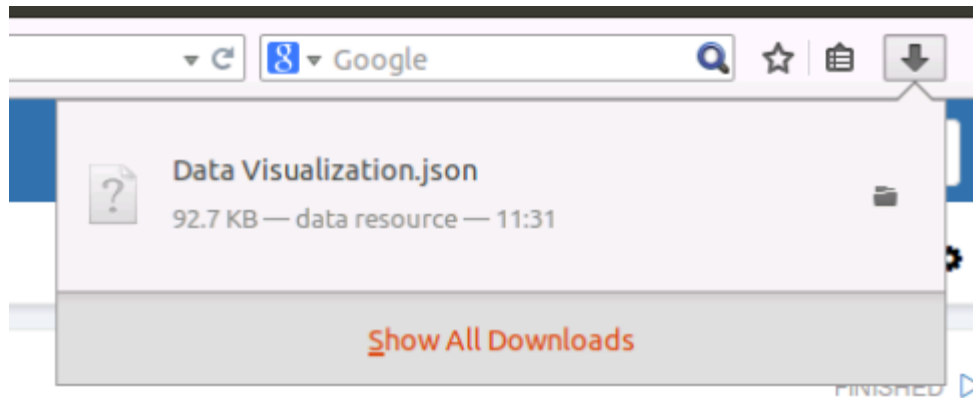
- o. We are now ready to prepare this note for sharing. Create a clone copy of this note named Data Visualization Clone. Also export a copy of the note.



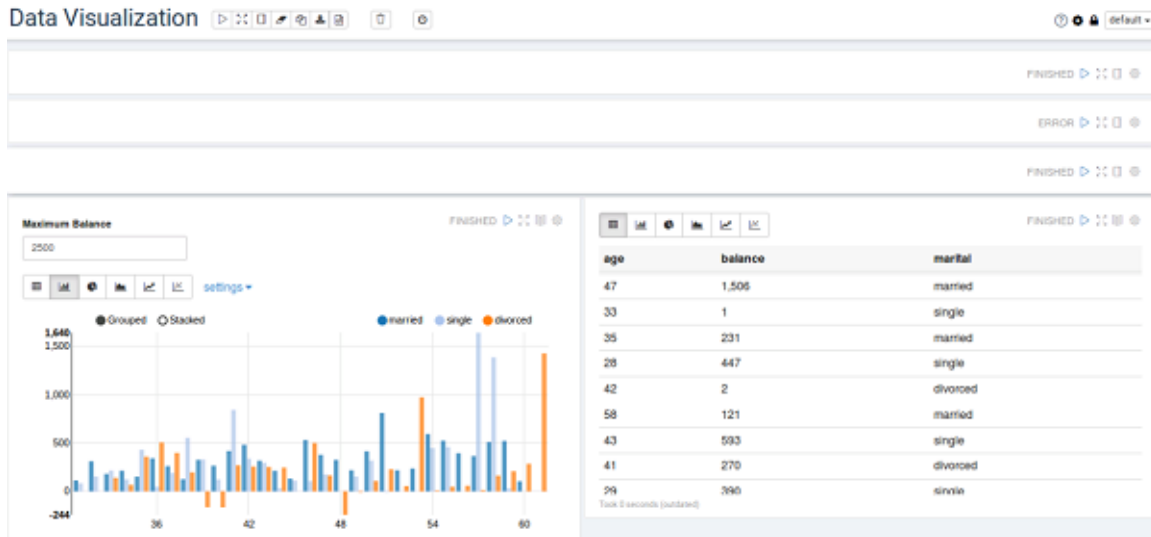
The screenshot shows a "Clone note" dialog box. The "Note Name" field contains the text "Data Visualization Clone". A "Clone Note" button is located at the bottom right of the dialog.



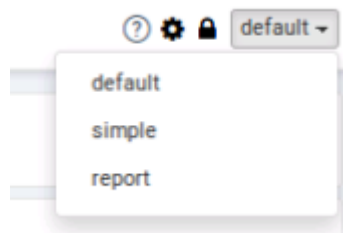
Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



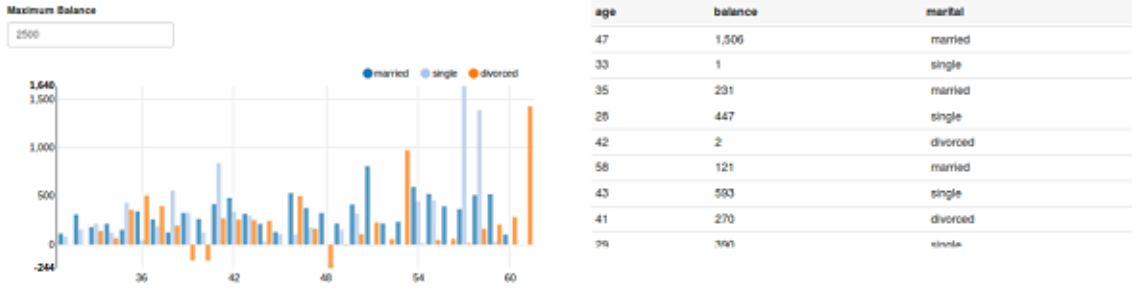
- p. On the Data Visualization note we are going to share, hide the code for all paragraphs. Then hide the output for every paragraph except for the two that are on the same line.



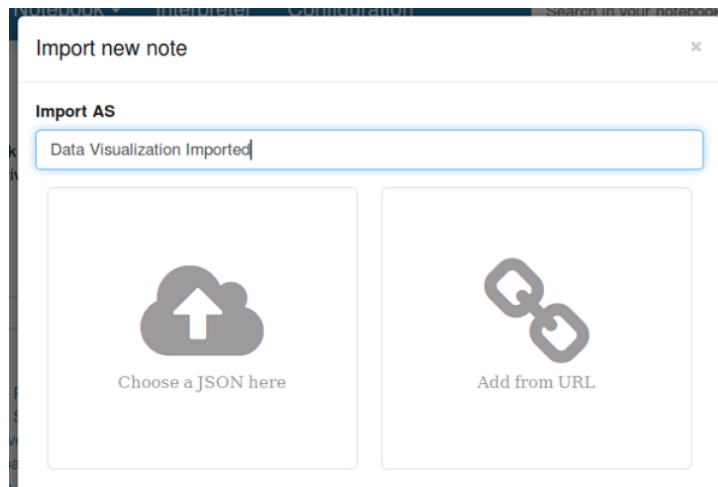
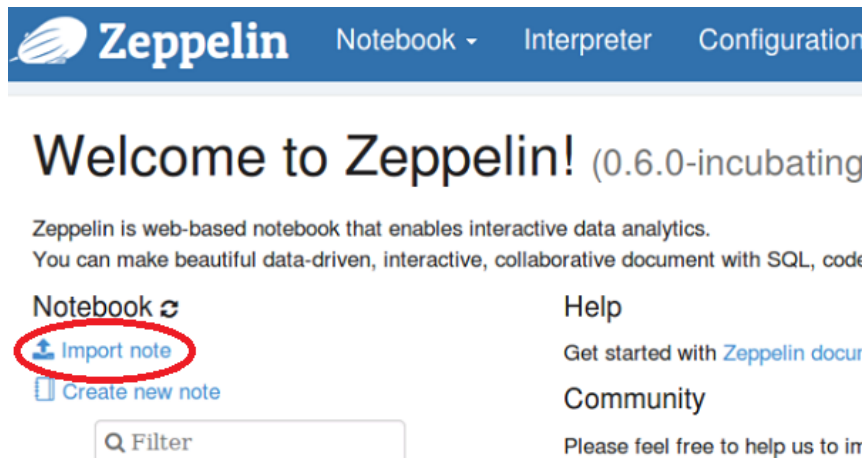
- q. Next, convert this from the default view to report view. Now the URL to this note is ready to share with your stakeholders.



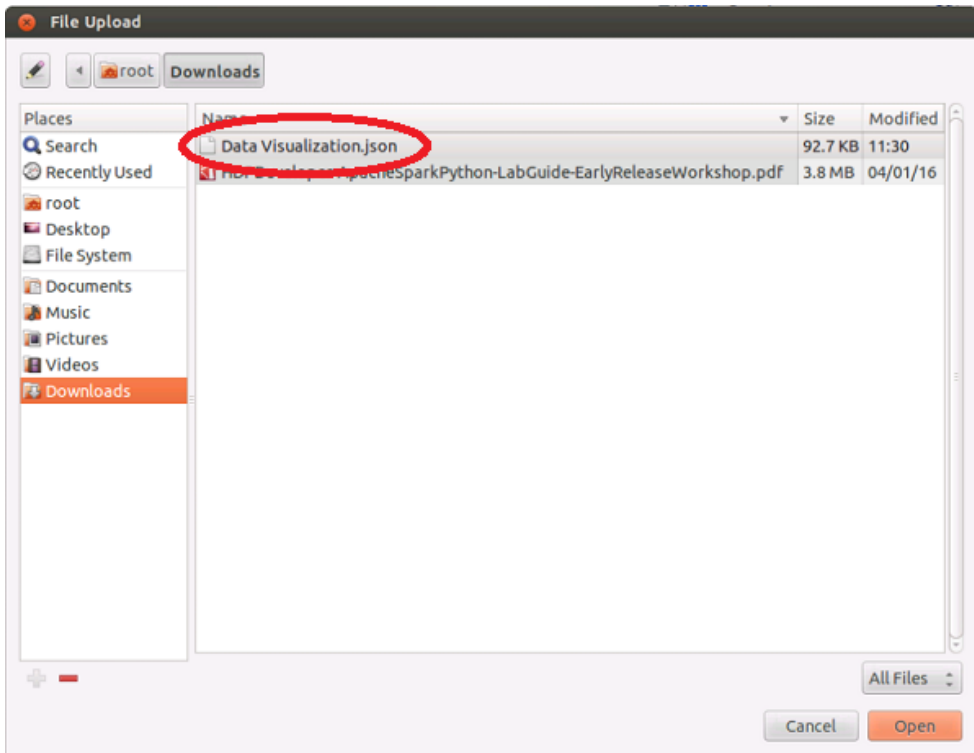
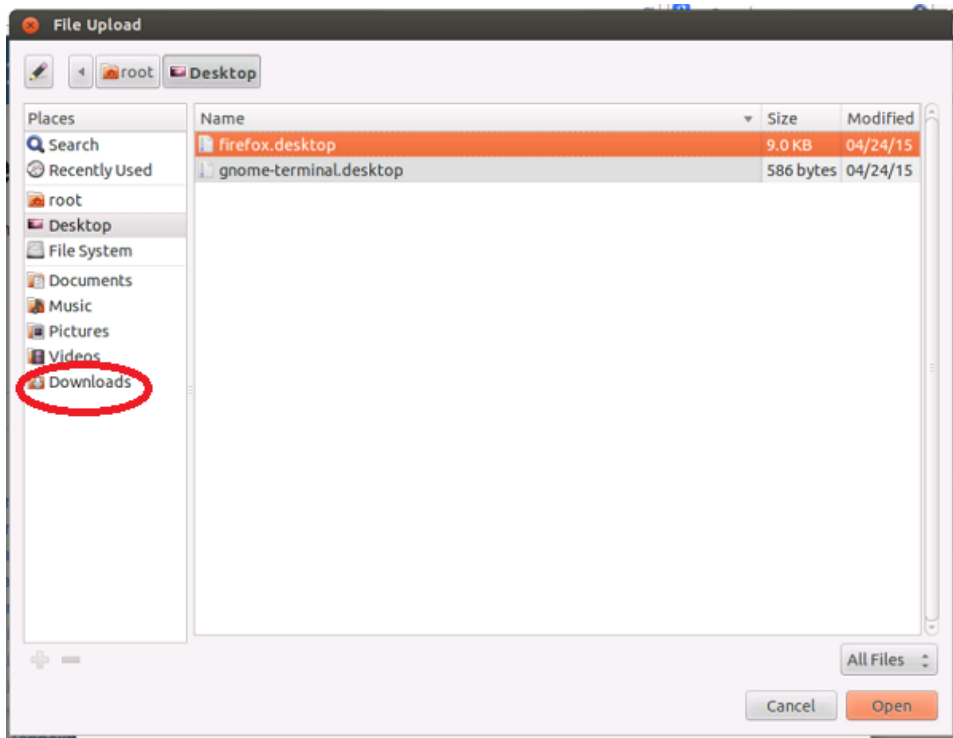
Data Visualization



- r. Import the copy of this note you made earlier and name the new note Data Visualization Imported. Confirm that the copy contains all original code and formatting.



Lab 10: Data Visualization, Reporting and Collaboration using Zeppelin (Scala)



- Data Visualization
- Data Visualization Clone
- Data Visualization Imported
- Hello World Tutorial

Data Visualization Imported

name	value
bankdataperm	false
table1hive	false

Maximum Balance:

```
sql
select * from bankdataperm
```

age	balance	marital
-----	---------	---------

Result

You have successfully created and manipulated Zeppelin visualizations, made them available for collaboration, and used Zeppelin to create a shareable report.

Lab 11: Job Monitoring (Scala)

About This Lab

Objective:

Monitor Spark jobs using the Spark Application UI

File Locations:

NA

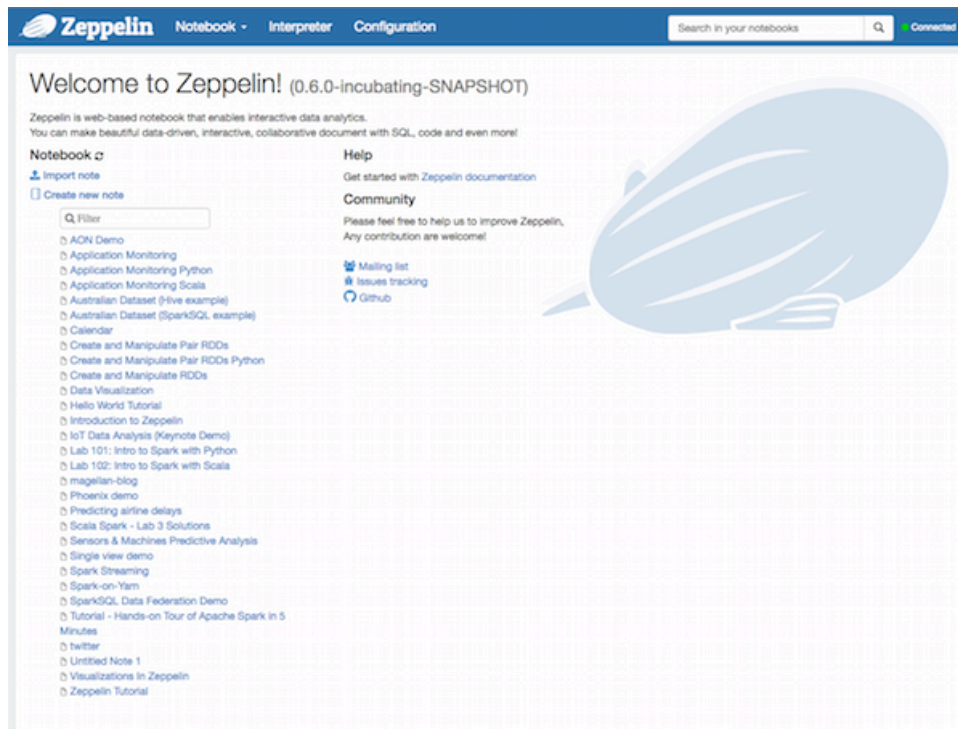
Successful Outcome:

Monitor Spark jobs.

Lab Steps

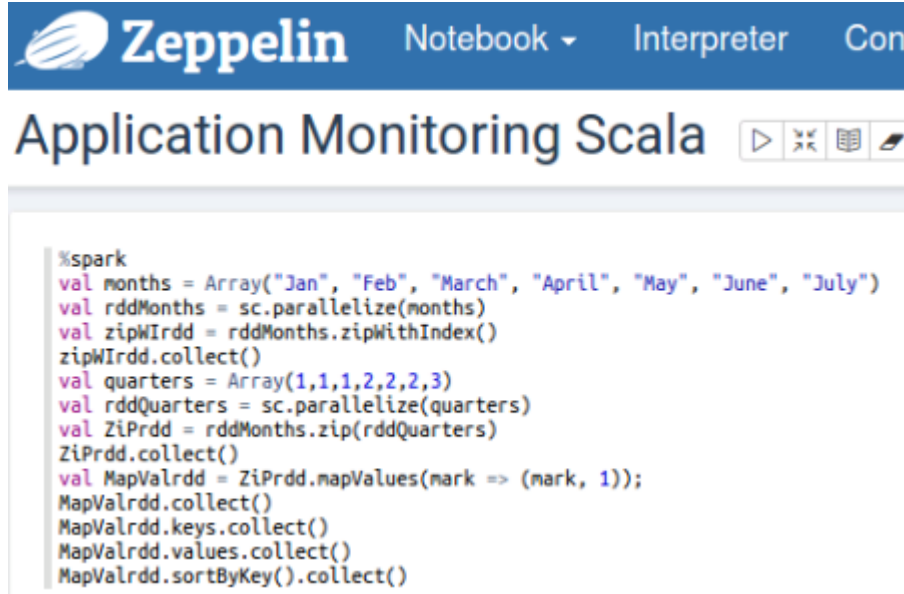
Perform the following steps:

1. Monitor a core RDD programming job.
 - a. Open the Firefox browser and access your Zeppelin notebook.
<http://sandbox:9995/>



- b. From the home page, select the Application Monitoring Scala Note. This note has prebuilt code that we will run to generate Spark job activity.

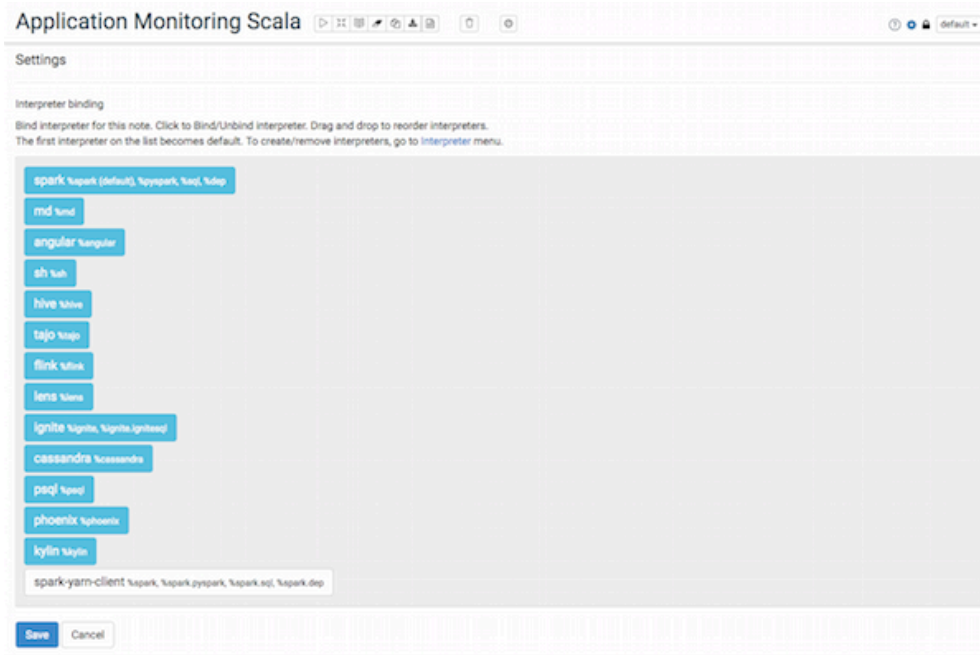
 [Application Monitoring Scala](#)



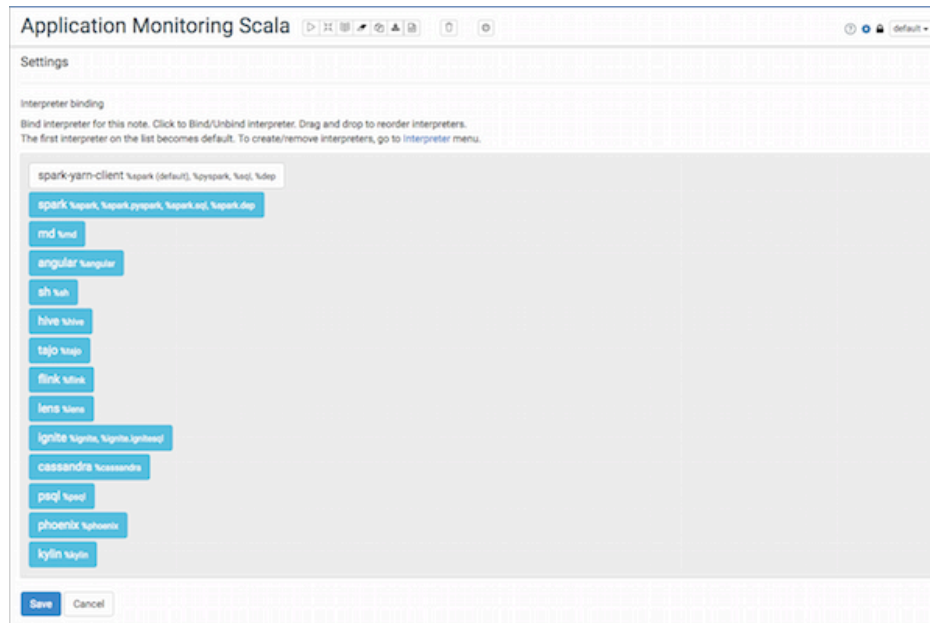
- c. At the top right click on the gear icon to change interpreter binding. Your administrator has enabled an interpreter called “**spark yarn-client**” which is configured for the HDP cluster you are using. Drag it to the top of the list of interpreters, and click the Save button.



Lab 11: Job Monitoring (Scala)



The screenshot shows the 'Application Monitoring Scala' settings page. Under the 'Interpreter binding' section, there is a list of interpreters. The first item is 'spark %spark (default), %yyspark, %scd, %ddp'. Below it are 'md %md', 'angular %angular', 'sh %sh', 'hive %hive', 'tajo %tajo', 'flink %flink', 'lens %lens', 'ignite %ignite, %ignite.ignitecd', 'cassandra %cassandra', 'psql %psql', 'phoenix %phoenix', and 'kylin %kylin'. At the bottom of the list is 'spark-yarn-client %spark, %spark.yyspark, %spark.scj, %spark.dcp'. There are 'Save' and 'Cancel' buttons at the bottom left of the list area.

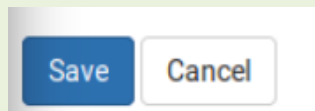


This screenshot is similar to the one above, but the 'spark %spark (default), %yyspark, %scd, %ddp' item is now highlighted in blue, indicating it is the selected default interpreter. The 'Save' and 'Cancel' buttons remain at the bottom left.



NOTE:

The first interpreter on the list is treated as the default interpreter. Scroll down to find the Save button.



Lab 11: Job Monitoring (Scala)

- d. Now run the code by hitting Play button  or by pressing **Shift + Enter**.



NOTE:

The below code is for reference purposes and has already been placed in the note.

```
%spark
val months = Array("Jan", "Feb", "March", "April", "May", "June", "July")
val rddMonths = sc.parallelize(months)
val zipWIrdd = rddMonths.zipWithIndex()
zipWIrdd.collect()
val quarters = Array(1,1,1,2,2,2,3)
val rddQuarters = sc.parallelize(quarters)
val ZiPrdd = rddMonths.zip(rddQuarters)
ZiPrdd.collect()
val MapValrdd = ZiPrdd.mapValues(mark => (mark, 1));
MapValrdd.collect()
MapValrdd.keys.collect()
MapValrdd.values.collect()
MapValrdd.sortByKey().collect()
```



The screenshot shows the Application Monitoring Scala interface. The code is the same as in the previous block. The output shows the following results:

```
months: Array[String] = Array(Jan, Feb, March, April, May, June, July)
rddMonths: org.apache.spark.rdd.RDD[String] = ParallelCollectionRDD[30] at parallelize at <console>:31
zipWIrdd: org.apache.spark.rdd.RDD[(String, Long)] = ZippedWithIndexRDD[31] at zipWithIndex at <console>:33
res0: Array[(String, Long)] = Array((Jan,0), (Feb,1), (March,2), (April,3), (May,4), (June,5), (July,6))
quarters: Array[Int] = Array(1, 1, 1, 2, 2, 2, 3)
rddQuarters: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[32] at parallelize at <console>:31
ZiPrdd: org.apache.spark.rdd.RDD[(String, Int)] = ZippedPartitionsRDD[33] at zip at <console>:37
res1: Array[(String, Int)] = Array((Jan,1), (Feb,1), (March,1), (April,2), (May,2), (June,2), (July,3))
MapValrdd: org.apache.spark.rdd.RDD[(String, (Int, Int))] = MapPartitionsRDD[34] at mapValues at <console>:39
res2: Array[(String, (Int, Int))] = Array((Jan,(1,1)), (Feb,(1,1)), (March,(1,1)), (April,(2,1)), (May,(2,1)), (June,(2,1)), (July,(3,1)))
res3: Array[String] = Array(Jan, Feb, March, April, May, June, July)
res4: Array[(Int, Int)] = Array((1,1), (1,1), (1,1), (2,1), (2,1), (2,1), (3,1))
res5: Array[(String, (Int, Int))] = Array((April,(2,1)), (Feb,(1,1)), (Jan,(1,1)), (July,(3,1)), (June,(2,1)), (March,(1,1)), (May,(2,1)))
```

- e. Open a new tab on the Firefox browser and enter the following URL to view the Spark Application UI:

<http://sandbox:4040/>





NOTE:

<http://sandbox:4040/> will work only once the job is submitted.

Job Id (Job Group)	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
8 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-13	2016/06/06 07:07:16	0.4 s	2/2	4/4
7 (zeppelin-20151218-011248_1336183271)	Zeppelin sortByKey at <string>-13	2016/06/06 07:07:16	0.1 s	1/1	2/2
6 (zeppelin-20151218-011248_1336183271)	Zeppelin sortByKey at <string>-13	2016/06/06 07:07:16	0.1 s	1/1	2/2
5 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-12	2016/06/06 07:07:16	87 ms	1/1	2/2
4 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-11	2016/06/06 07:07:16	92 ms	1/1	2/2
3 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-10	2016/06/06 07:07:16	92 ms	1/1	2/2
2 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-9	2016/06/06 07:07:15	75 ms	1/1	2/2
1 (zeppelin-20151218-011248_1336183271)	Zeppelin	2016/06/06 07:07:15	0.1 s	1/1	2/2



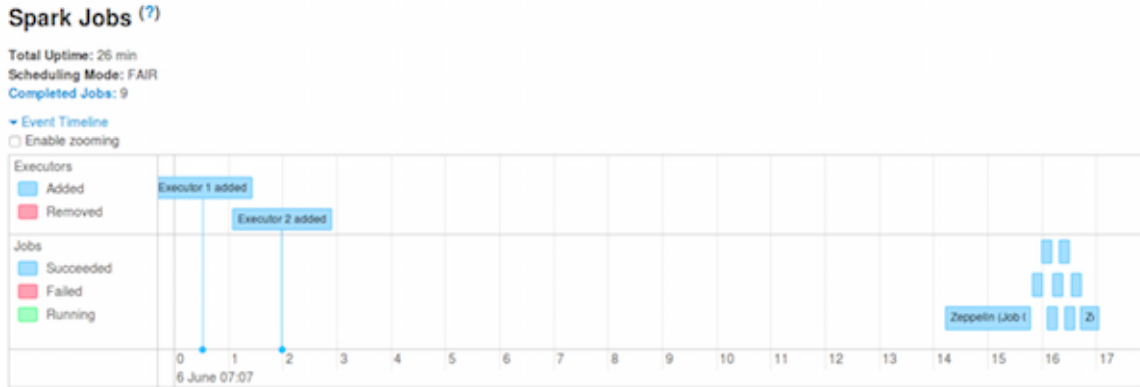
NOTE:

the URL <http://sandbox:4040/> has been redirected to http://sandbox:8088/proxy/application_ID. Port 8088 belongs to the job history server for various applications that run on YARN. Here our application is “Zeppelin application UI” as noted in the top-right corner of the window.

Zeppelin application UI

f. SPARK APPLICATION UI SCAVENGER HUNT!

Look at the various aspects of the jobs that were run as part of the code being executed in the step above. Try to locate the following screens (the details of your environment may differ from the details shown):



Completed Jobs (9)

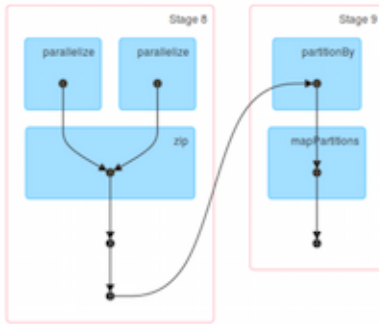
Completed Jobs (9)

Job Id (Job Group)	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
8 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-13	2016/06/06 07:07:16	0.4 s	2/2	4/4
7 (zeppelin-20151218-011248_1336183271)	Zeppelin sortByKey at <string>-13	2016/06/06 07:07:16	0.1 s	1/1	2/2
6 (zeppelin-20151218-011248_1336183271)	Zeppelin sortByKey at <string>-13	2016/06/06 07:07:16	0.1 s	1/1	2/2
5 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-12	2016/06/06 07:07:16	87 ms	1/1	2/2
4 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-11	2016/06/06 07:07:16	92 ms	1/1	2/2
3 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-10	2016/06/06 07:07:16	92 ms	1/1	2/2
2 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-8	2016/06/06 07:07:15	75 ms	1/1	2/2
1 (zeppelin-20151218-011248_1336183271)	Zeppelin collect at <string>-4	2016/06/06 07:07:15	0.1 s	1/1	2/2
0 (zeppelin-20151218-011248_1336183271)	Zeppelin zipWithIndex at <string>-3	2016/06/06 07:07:14	2 s	1/1	2/2

Details for Job 8

Status: SUCCEEDED
Job Group: zeppelin-20151218-011248_1336183271
Completed Stages: 2

- Event Timeline
- DAG Visualization



Completed Stages (2)

Lab 11: Job Monitoring (Scala)

Completed Stages (2)

Stage Id	Pool Name	Description	Submitted	Duration	Tasks: Succeeded/Total	Input	Output	Shuffle Read	Shuffle Write
9	default	Zeppelin collect at <string>.13	2016/06/06 07:07:16	0.2 s	2/2			594.0 B	
8	default	Zeppelin sortByKey at <string>.13	2016/06/06 07:07:16	0.1 s	2/2				594.0 B

Spark 1.6.0 Jobs Stages Storage Environment Executors SQL Zeppelin application UI

Stages for All Jobs

Completed Stages: 10

2 Fair Scheduler Pools

Pool Name	Minimum Share	Pool Weight	Active Stages	Running Tasks	SchedulingMode
default	0	1	0	0	FIFO
fair	0	1	0	0	FAIR

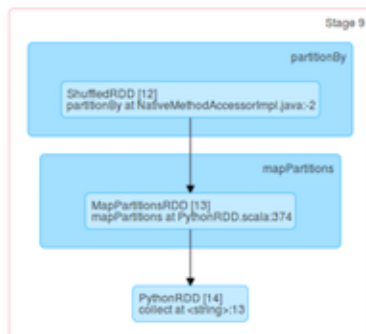
Completed Stages (10)

Stage Id	Pool Name	Description	Submitted	Duration	Tasks: Succeeded/Total	Input	Output	Shuffle Read	Shuffle Write
9	default	Zeppelin collect at <string>.13	2016/06/06 07:07:16	0.2 s	2/2			594.0 B	
8	default	Zeppelin sortByKey at <string>.13	2016/06/06 07:07:16	0.1 s	2/2				594.0 B
7	default	Zeppelin sortByKey at <string>.13	2016/06/06 07:07:16	0.1 s	2/2				
6	default	Zeppelin sortByKey at <string>.13	2016/06/06 07:07:16	0.1 s	2/2				
5	default	Zeppelin collect at <string>.12	2016/06/06 07:07:16	77 ms	2/2				
4	default	Zeppelin collect at <string>.11	2016/06/06 07:07:16	84 ms	2/2				

Details for Stage 9 (Attempt 0)

Total Time Across All Tasks: 0.2 s
 Locality Level Summary: Node local: 2
 Shuffle Read: 594.0 B / 4

▼ DAG Visualization



When you get to the Show Additional Metrics link, try reading about and selecting additional metrics and view the information they provide. How might this be useful in troubleshooting application performance problems?

Lab 11: Job Monitoring (Scala)

▼ Show Additional Metrics

- (De)select All
- Scheduler Delay
- Task Deserialization Time
- Shuffle Read Blocked Time
- Shuffle Remote Reads
- Result Serialization Time
- Getting Result Time
- Peak Execution Memory

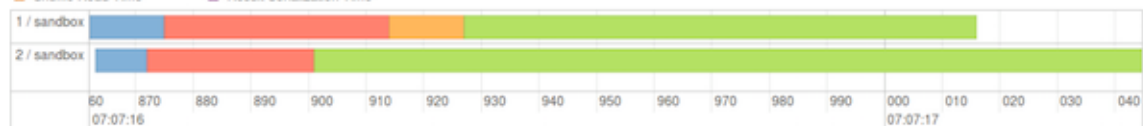
Summary Metrics for 2 Completed Tasks

Metric	Min	25th percentile	Median	75th percentile	Max
Duration	0.1 s	0.1 s	0.1 s	0.1 s	0.1 s
GC Time	0 ms	0 ms	0 ms	0 ms	0 ms
Shuffle Read Size / Records	291.0 B / 2	291.0 B / 2	303.0 B / 2	303.0 B / 2	303.0 B / 2

▼ Event Timeline

Enable zooming

- Scheduler Delay
- Task Deserialization Time
- Shuffle Read Time
- Executor Computing Time
- Shuffle Write Time
- Result Serialization Time
- Getting Result Time



2. Monitor a Spark Streaming job.

- a. Open a terminal window and SSH into sandbox.

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
```

- b. Start a new REPL specifying the local machine as the master and allocate two cores for the streaming application.

```
# spark-shell --master local[2]
```

```
[root@sandbox ~]# spark-shell --master local[2]
```

- c. Set the log level to ERROR to avoid screen clutter while running the streaming application.

```
scala> sc.setLogLevel("ERROR")
```

```
scala> sc.setLogLevel("ERROR")
```

- d. Import the streaming library.

```
scala> import org.apache.spark.streaming._
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.StreamingContext@3e216fc9
```

- e. Create a streaming context with a five-second batch duration.

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
```

```
scala> val sscFive = new StreamingContext(sc, Seconds(5))
sscFive: org.apache.spark.streaming.StreamingContext = org.apache.spark.streaming.StreamingContext@3e216fc9
```

- f. Create a DStream using `socketTestStream()` to the system named “sandbox” on port 9999.

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)
```

```
scala> val inputDS = sscFive.socketTextStream("sandbox", 9999)
inputDS: org.apache.spark.streaming.dstream.ReceiverInputDStream[String] = org.apache.spark.streaming.dstream.SocketInputDStream@4caeeab6
```

- g. Print out the output to the terminal window.

```
scala> inputDS.print()
```

```
scala> inputDS.print()
```

- h. Start the streaming application. Note that only new files will be streamed, so any files that existed at application launch will not be streamed.

```
scala> sscFive.start()
```

```
scala> sscFive.start()
```



NOTE:

An error will appear when the application starts because the application is waiting for an input connection.

Lab 11: Job Monitoring (Scala)

```
at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
16/05/30 16:20:58 ERROR ReceiverTracker: Deregistered receiver for stream 0: Restarting receiver with delay 2000ms: Error connecting to sandbox:999 - java.net.ConnectException: Connection refused
    at java.net.PlainSocketImpl.socketConnect(Native Method)
    at java.net.AbstractPlainSocketImpl.doConnect(AbstractPlainSocketImpl.java:345)
    at java.net.AbstractPlainSocketImpl.connectToAddress(AbstractPlainSocketImpl.java:206)
    at java.net.AbstractPlainSocketImpl.connect(AbstractPlainSocketImpl.java:188)
    at java.net.SocksSocketImpl.connect(SocksSocketImpl.java:392)
    at java.net.Socket.connect(Socket.java:589)
    at java.net.Socket.connect(Socket.java:538)
    at java.net.Socket.<init>(Socket.java:434)
    at java.net.Socket.<init>(Socket.java:211)
    at org.apache.spark.streaming.dstream.SocketReceiver.receive(SocketInputDStream.scala:73)
    at org.apache.spark.streaming.dstream.SocketReceiver$$anon$2.run(SocketInputDStream.scala:59)
```

- i. In a second terminal window SSH to sandbox and use the netcat utility to create a connection to port 9999.

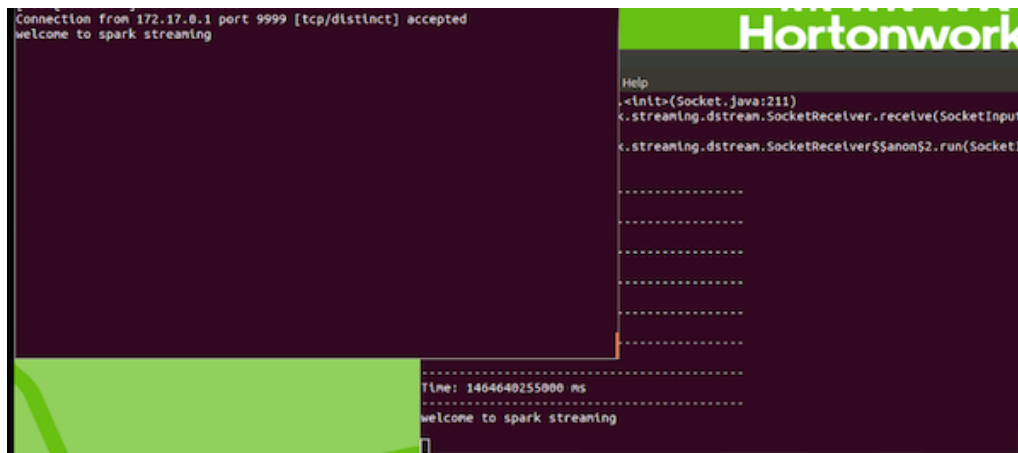
```
# ssh sandbox
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

```
# nc -lkv 9999
```

```
[root@sandbox ~]# nc -lkv 9999
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
```

- j. Start typing words separated by space, hit **Enter** occasionally to submit them. Observe what happens in the streaming terminal window a few seconds after hitting **Enter**.



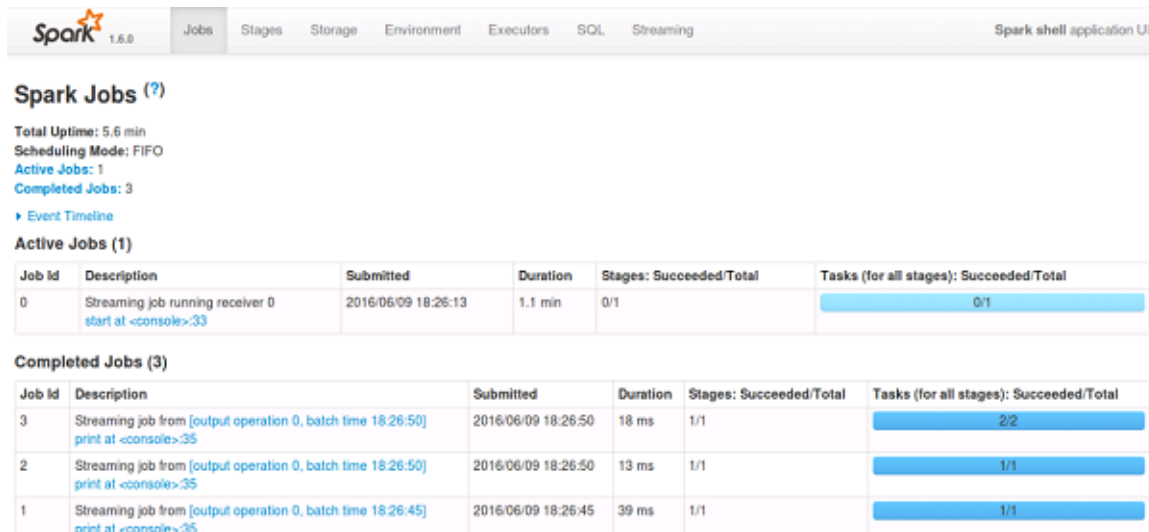
```
Connection from 172.17.0.1 port 9999 [tcp/distinct] accepted
welcome to spark streaming
-----
Time: 1464640255000 ns
welcome to spark streaming
```

Lab 11: Job Monitoring (Scala)

- k. Once you observe data being streamed on-screen in the first terminal window, use **Ctrl + C** (or **Cmd + C** if using a Mac) to exit netcat in the second terminal window.

```
[root@sandbox data]# nc -l kv 9999
Hello world
This is an example of streaming data
Random words random words
^C
[root@sandbox data]#
```

- l. Since this is a new SparkContext instance, a new Spark Applications UI should now be available. Open a new FireFox tab and browse to the Streaming Application UI URL from before, but replace port 4040 with 4041:



The screenshot shows the Spark Applications UI for a streaming job. The top navigation bar includes 'Spark 1.6.0', 'Jobs', 'Stages', 'Storage', 'Environment', 'Executors', 'SQL', and 'Streaming'. The main content area is titled 'Spark Jobs (?)' and displays summary statistics: Total Uptime: 5,6 min, Scheduling Mode: FIFO, Active Jobs: 1, and Completed Jobs: 3. Below this, there are two tables: 'Active Jobs (1)' and 'Completed Jobs (3)'. The 'Active Jobs' table shows a single job with ID 0, description 'Streaming job running receiver 0 start at <console>-33', submitted at 2016/06/09 18:26:13, duration of 1.1 min, and 0/1 stages. The 'Completed Jobs' table shows three jobs with IDs 3, 2, and 1, all with a duration of approximately 18-39 ms and 1/1 stages. Each job's progress is visualized with a blue progress bar.

Job Id	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
0	Streaming job running receiver 0 start at <console>-33	2016/06/09 18:26:13	1.1 min	0/1	0/1

Job Id	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
3	Streaming job from [output operation 0, batch time 18:26:50] print at <console>-35	2016/06/09 18:26:50	18 ms	1/1	2/2
2	Streaming job from [output operation 0, batch time 18:26:50] print at <console>-35	2016/06/09 18:26:50	13 ms	1/1	1/1
1	Streaming job from [output operation 0, batch time 18:26:45] print at <console>-35	2016/06/09 18:26:45	39 ms	1/1	1/1

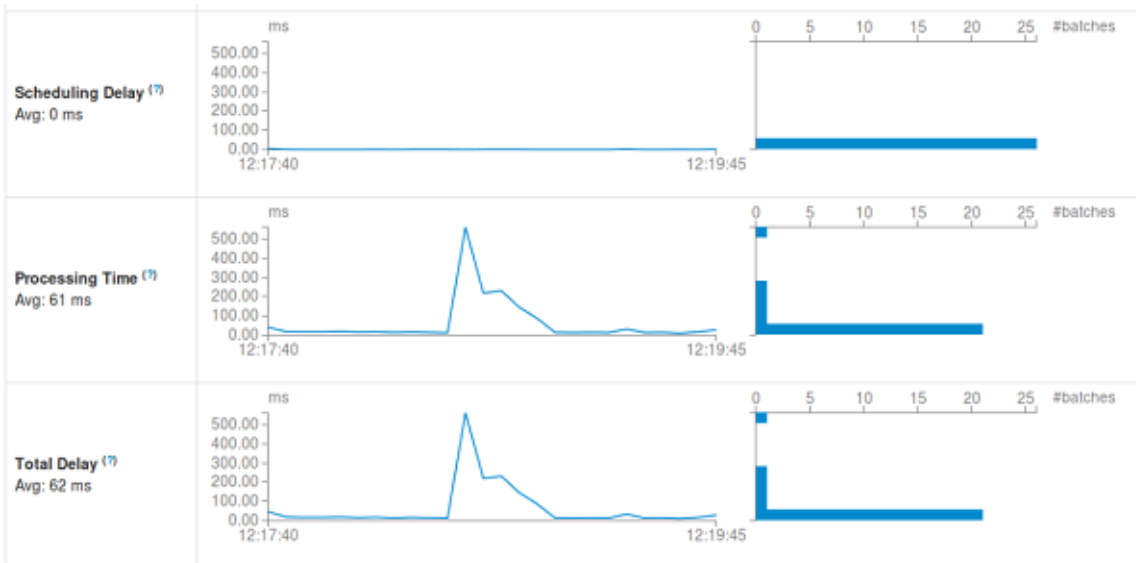
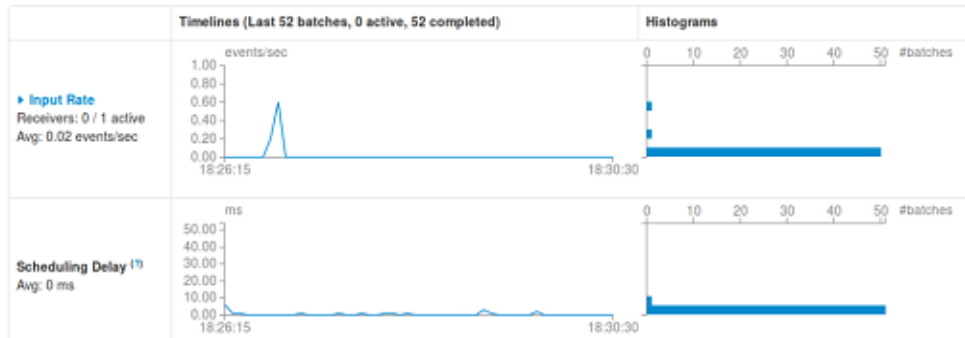
m. SCAVENGER HUNT PART II

Look at the various aspects of the streaming jobs that were run as part of the code being executed in the steps above. Try to locate the following screens (the details of your environment may differ from the details shown):

Lab 11: Job Monitoring (Scala)

Streaming Statistics

Running batches of 5 seconds for 7 minutes 38 seconds since 2016/06/09 18:22:51 (52 completed batches, 4 records)



Active Batches (0)

Batch Time	Input Size	Scheduling Delay ^(?)	Processing Time ^(?)	Output Ops: Succeeded/Total	Status
------------	------------	---------------------------------	--------------------------------	-----------------------------	--------

Completed Batches (last 26 out of 26)

Batch Time	Input Size	Scheduling Delay ^(?)	Processing Time ^(?)	Total Delay ^(?)	Output Ops: Succeeded/Total
2016/06/09 12:19:45	0 events	2 ms	27 ms	29 ms	1/1
2016/06/09 12:19:40	0 events	0 ms	17 ms	17 ms	1/1
2016/06/09 12:19:35	0 events	1 ms	10 ms	11 ms	1/1
2016/06/09 12:19:30	0 events	0 ms	14 ms	14 ms	1/1
2016/06/09 12:19:25	0 events	0 ms	13 ms	13 ms	1/1
2016/06/09 12:19:20	0 events	3 ms	31 ms	34 ms	1/1
2016/06/09 12:19:15	0 events	0 ms	13 ms	13 ms	1/1
2016/06/09 12:19:10	0 events	0 ms	14 ms	14 ms	1/1
2016/06/09 12:19:05	0 events	0 ms	13 ms	13 ms	1/1
2016/06/09 12:19:00	0 events	0 ms	14 ms	14 ms	1/1
2016/06/09 12:18:55	0 events	0 ms	87 ms	87 ms	1/1

Lab 11: Job Monitoring (Scala)

Spark 1.6.0 Jobs Stages Storage Environment Executors SQL Streaming Spark shell application UI

Details of batch at 2016/06/09 18:30:30

Batch Duration: 5 s
Input data size: 0 records
Scheduling delay: 0 ms
Processing time: 1 ms
Total delay: 1 ms

Output Op Id	Description	Duration	Status	Job Id	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total	Error
0	print at <console>35	1 ms	Succeeded	-	-	-	-	-

▼ Input Rate
Receivers: 1 / 1 active
Avg: 0.02 events/sec

Status	Executor ID / Host	Last Error Time	Last Error Message
ACTIVE	driver / localhost	-	-

SocketReceiver-0
Avg: 0.02 events/sec

- n. When you have located all of the required sections, go back to the first terminal window, stop the stream and exit the REPL. If the stream refreshes while you are typing, that will not affect the input. Simply continue to type the command and press enter.

```
sc.stop()  
exit()
```

Result

You have successfully monitored Spark core programming and Spark Streaming jobs using the Spark Application UI.

Lab 12: Performance Tuning (Scala)

About This Lab

Objective:

Practice performance tuning techniques

File Locations:

/home/zeppelin/spark/data/

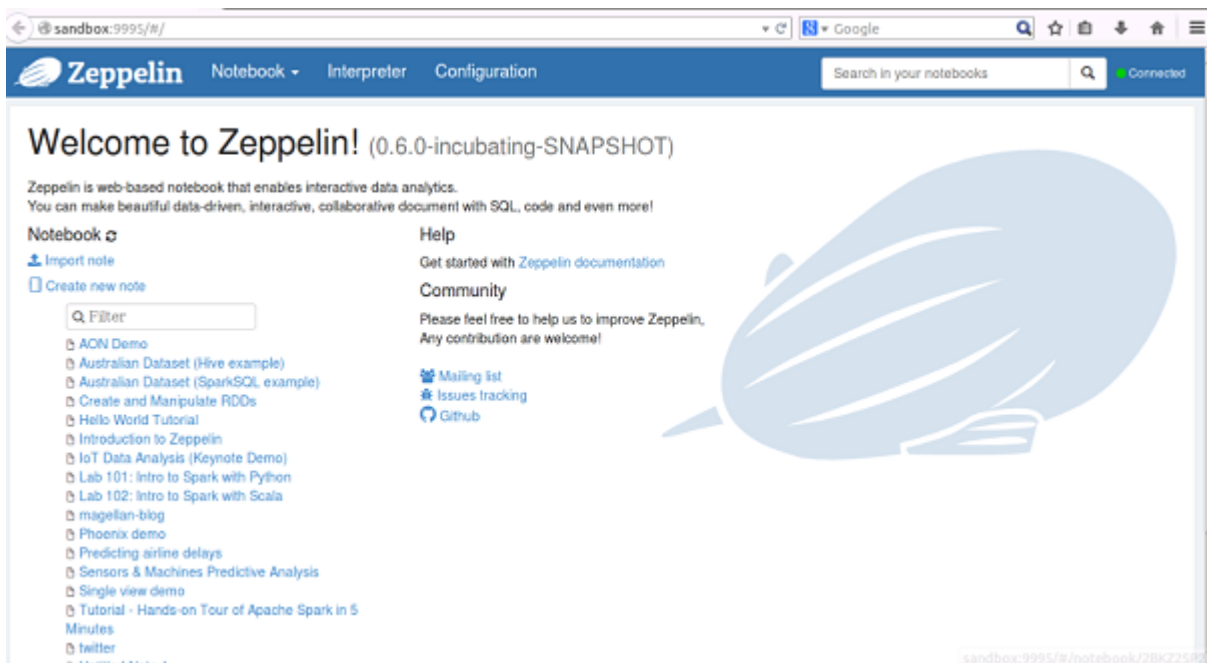
Successful Outcome:

Code performance tuning techniques from the lesson

Lab Steps

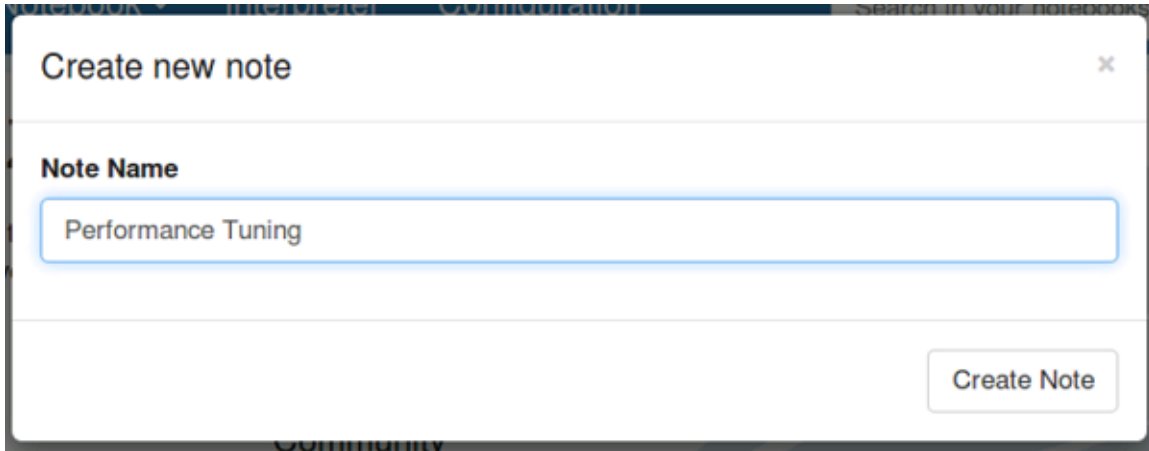
Perform the following steps:

- 1 . Practice using performance tuning techniques.
 - a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.
<http://sandbox:9995/>

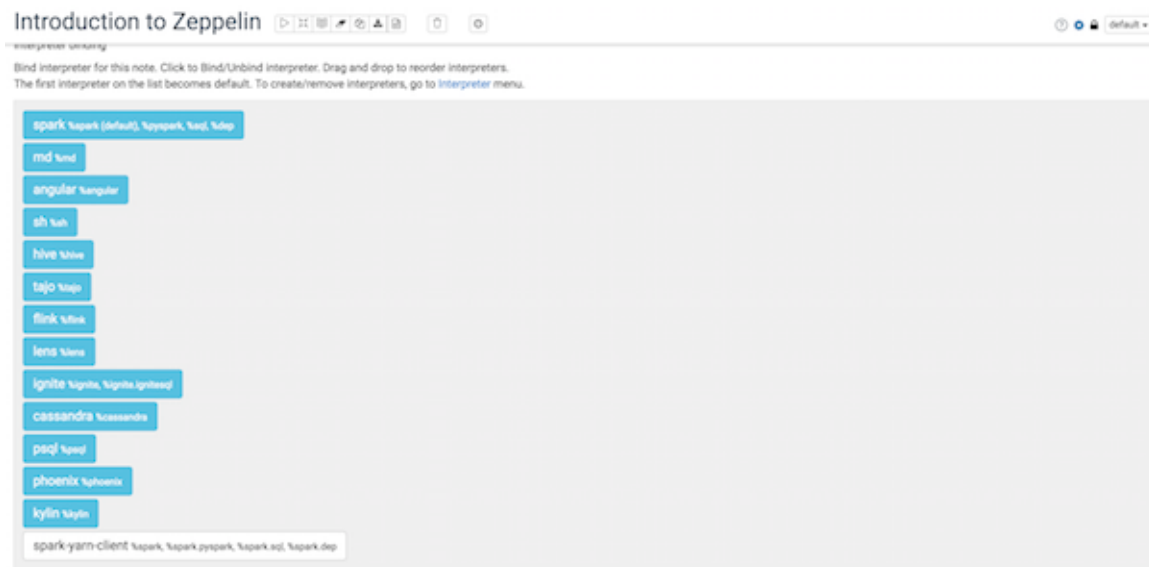


Lab 12: Performance Tuning (Scala)

- b. Click on Create new note. Name this note Performance Tuning.

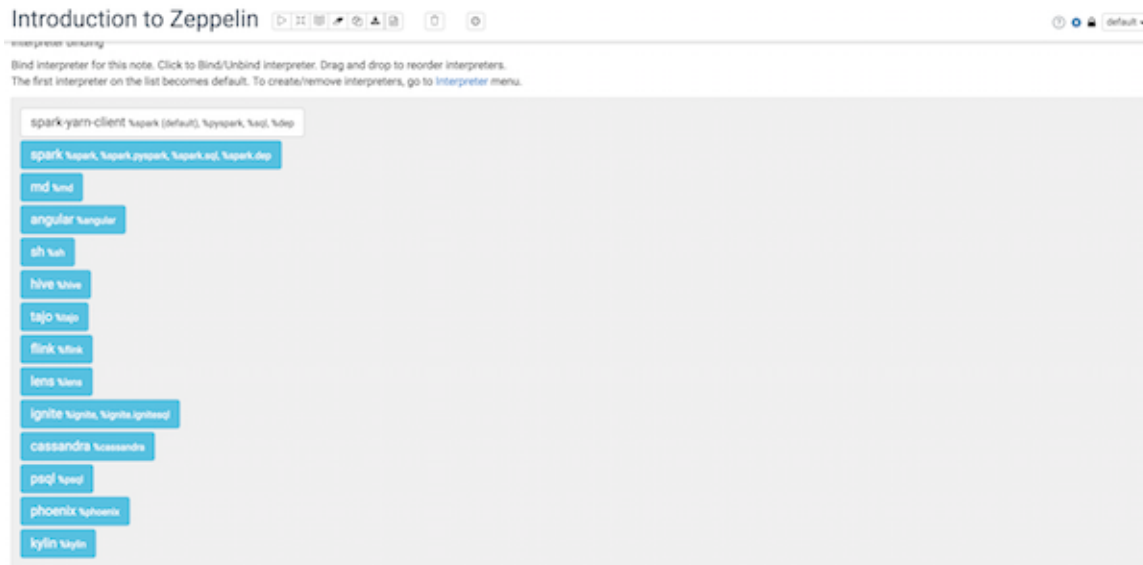


- c. At the top right click on the gear icon to change interpreter binding.

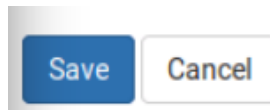


Lab 12: Performance Tuning (Scala)

Drag the spark-yarn-client to the top and click save.



The first interpreter on the list becomes default.



- d. Create an RDD named `rdd1` that contains a list of numbers one through nine, then back down to one again (17 elements total) and set it to eight partitions. Use `print` to confirm the RDD was created successfully.

```
val rdd1 = sc.parallelize(Seq(1, 2, 3, 4, 5, 6, 7, 8, 9, 8, 7, 6, 5, 4, 3, 2, 1), 8)

rdd1.collect()
```

```
val rdd1 = sc.parallelize(Seq(1, 2, 3, 4, 5, 6, 7, 8, 9, 8, 7, 6, 5, 4, 3, 2, 1), 8)
rdd1.collect()

rdd1: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[711] at parallelize at <console>:29
res44: Array[Int] = Array(1, 2, 3, 4, 5, 6, 7, 8, 9, 8, 7, 6, 5, 4, 3, 2, 1)
```

- e. View the default parallelism settings for your environment, and then verify that `rdd1` was partitioned with eight partitions instead of the default number.

```
sc.defaultParallelism
rdd1.getNumPartitions
```

```
sc.defaultParallelism
rdd1.getNumPartitions]
res53: Int = 2
res54: Int = 8
```

- f. Create an RDD named `rdd2` that is a copy of `rdd1` but uses only four partitions. Verify that `rdd2` has only four partitions.

```
val rdd2 = rdd1.coalesce(4)
rdd2.getNumPartitions
```

```
val rdd2 = rdd1.coalesce(4)
rdd2.getNumPartitions]
rdd2: org.apache.spark.rdd.RDD[Int] = CoalescedRDD[712] at coalesce at <console>:31
res56: Int = 4
```

- g. Create an RDD named `rdd3` that is a copy of `rdd2` but expands the number of partitions from four to six. Verify that `rdd3` has six partitions.

```
val rdd3 = rdd2.repartition(6)
rdd3.getNumPartitions
```

```
val rdd3 = rdd2.repartition(6)
rdd3.getNumPartitions]
rdd3: org.apache.spark.rdd.RDD[Int] = MapPartitionsRDD[716] at repartition at <console>:33
res58: Int = 6
- - -
```

- h. Create an RDD named `rdd4` that contains a larger set of data by combining `rdd3`, `rdd2`, and `rdd1`. The view this list of 51 numbers.

```
val rdd4 = rdd3.union(rdd2.union(rdd1))
rdd4.collect()
```

Lab 12: Performance Tuning (Scala)

```
val rdd4 = rdd3.union(rdd2.union(rdd1))
rdd4.collect()

rdd4: org.apache.spark.rdd.RDD[Int] = UnionRDD[719] at union at <console>:35
res66: Array[Int] = Array(8, 2, 7, 8, 1, 7, 1, 2, 6, 3, 5, 4, 4, 5, 6, 9, 3, 1, 2, 3, 4, 5, 6, 7, 8, 9, 8, 7, 6, 5, 4, 3, 2, 1)
```

- i. Create an RDD named `rdd5` that turns this list into a Pair RDD using the existing numbers as keys and assign each key a value of one. View `rdd5` to confirm successful operation.

```
val rdd5 = rdd4.map(x => (x, 1))
rdd5.collect()
```

```
val rdd5 = rdd4.map(x => (x, 1))
rdd5.collect()

rdd5: org.apache.spark.rdd.RDD[(Int, Int)] = MapPartitionsRDD[720] at map at <console>:37
res62: Array[(Int, Int)] = Array((7,1), (8,1), (2,1), (1,1), (7,1), (1,1), (8,1), (2,1), (6,1), (3,1), (5,1), (4,1), (4,1), (5,1), (6,1), (9,1), (3,1), (1,1), (2,1), (3,1), (4,1), (5,1), (6,1), (7,1), (8,1), (9,1), (8,1), (7,1), (6,1), (5,1), (4,1), (3,1), (2,1), (1,1), (1,1), (2,1), (3,1), (4,1), (5,1), (6,1), (7,1), (8,1), (9,1), (8,1), (7,1), (6,1), (5,1), (4,1), (3,1), (2,1), (1,1))
```

- j. Create an RDD named `rdd6` that uses `partitionBy()` to create eight hashed partitions from `rdd5`. View `rdd6` to confirm successful operation.

```
import org.apache.spark.HashPartitioner

val rdd6 = rdd5.partitionBy(new HashPartitioner(8))
rdd6.collect()
```

```
import org.apache.spark.HashPartitioner
val rdd6 = rdd5.partitionBy(new HashPartitioner(8))
rdd6.collect()

import org.apache.spark.HashPartitioner
rdd6: org.apache.spark.rdd.RDD[(Int, Int)] = ShuffledRDD[721] at partitionBy at <console>:40
res64: Array[(Int, Int)] = Array((8,1), (8,1), (8,1), (8,1), (8,1), (8,1), (1,1), (1,1), (1,1), (9,1), (9,1), (1,1), (1,1), (9,1), (1,1), (2,1), (2,1), (2,1), (2,1), (2,1), (2,1), (3,1), (3,1), (3,1), (3,1), (3,1), (3,1), (3,1), (3,1), (4,1), (4,1), (4,1), (4,1), (4,1), (4,1), (5,1), (5,1), (5,1), (5,1), (5,1), (5,1), (6,1), (6,1), (6,1), (6,1), (6,1), (6,1), (7,1), (7,1), (7,1), (7,1), (7,1), (7,1), (7,1))
```

- k. Cache `rdd6` in memory so that it will be quickly available should we want to use the hash partitioning in a future operation.

```
rdd6.cache()
```

```
rdd6.cache()
```

```
res66: rdd6.type = ShuffledRDD[721] at partitionBy at <console>:40
```

Lab 12: Performance Tuning (Scala)

- I. Create a new RDD named `rdd7` that reduces `rdd6` by key. View the results and note the time it took to complete the operation.

```
val rdd7 = rdd6.reduceByKey((x,y) => x+y)
rdd7.collect()
```

```
val rdd7 = rdd6.reduceByKey((x,y) => x+y)
rdd7.collect()
```

```
rdd7: org.apache.spark.rdd.RDD[(Int, Int)] = MapPartitionsRDD[722] at reduceByKey at <console>:42
res68: Array[(Int, Int)] = Array((8,6), (1,6), (9,3), (2,6), (3,6), (4,6), (5,6), (6,6), (7,6))
```

```
took 5 seconds
```

- m. Create a directory named `checkperf` in your HDFS home directory, then configure it as your checkpoint directory for Spark applications.

```
%sh
hdfs dfs -mkdir checkperf
```

```
sc.setCheckpointDir("checkperf")
```

```
%sh
hdfs dfs -mkdir checkperf
```

```
sc.setCheckpointDir("checkperf")
```

- n. Checkpoint `rdd6` so that future operations can use it as the starting point for lineage-tracking purposes.

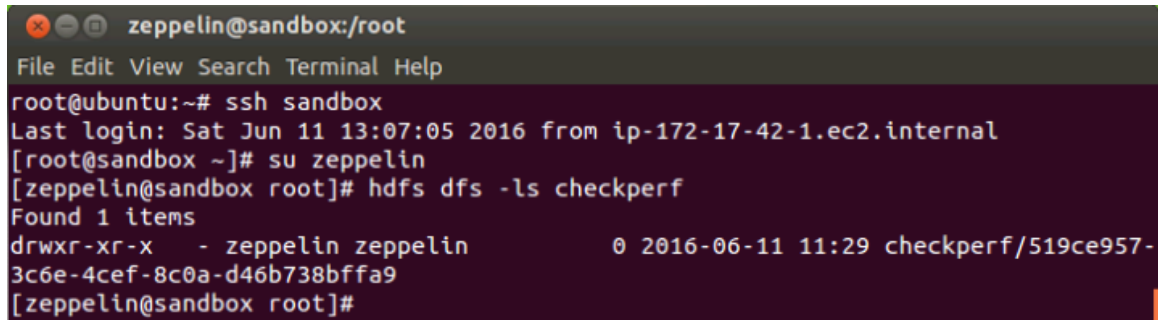
```
rdd6.checkpoint()
```

```
rdd6.checkpoint()
```


Lab 12: Performance Tuning (Scala)

- o. Open a terminal window and connect to sandbox using SSH. Switch to the zeppelin user. Then view the contents of the checkperf directory and confirm that a checkpoint file exists. Then exit the zeppelin user back to root.

```
# ssh sandbox
# su zeppelin
# hdfs dfs -ls checkperf
# exit
```



```
zppelin@sandbox:/root
File Edit View Search Terminal Help
root@ubuntu:~# ssh sandbox
Last login: Sat Jun 11 13:07:05 2016 from ip-172-17-42-1.ec2.internal
[root@sandbox ~]# su zeppelin
[zeppelin@sandbox root]# hdfs dfs -ls checkperf
Found 1 items
drwxr-xr-x  - zeppelin zeppelin          0 2016-06-11 11:29 checkperf/519ce957-
3c6e-4cef-8c0a-d46b738bffa9
[zeppelin@sandbox root]#
```

- p. Use broadcast variables to perform an operation. Code the following:
 1. Create a variable named oddNums that contains a list of odd numbers 1-9.
 2. Print the contents of rdd1 used at the beginning of the lab.
 3. Create a broadcast variable named filterOdd that contains the values in oddNums.
 4. Print the results of a filter operation where only numbers that appear in the filterOdd broadcast variable show up in the output.

```
val oddNums = (Array(1, 3, 5, 7, 9))
rdd1.collect()
val filterOdd = sc.broadcast(oddNums)
rdd1.filter(x => filterOdd.value contains x).collect()
```

```
val oddNums = (Array(1, 3, 5, 7, 9))  
rdd1.collect()  
val filterOdd = sc.broadcast(oddNums)  
rdd1.filter(x => filterOdd.value contains x).collect()
```

```
oddNums: Array[Int] = Array(1, 3, 5, 7, 9)  
res91: Array[Int] = Array(1, 2, 3, 4, 5, 6, 7, 8, 9, 8, 7, 6, 5, 4, 3, 2, 1)  
filterOdd: org.apache.spark.broadcast.Broadcast[Array[Int]] = Broadcast(295)  
res92: Array[Int] = Array(1, 3, 5, 7, 9, 7, 5, 3, 1)  
-- -- --
```

Result

You have used several of the performance tuning tools and practices discussed in the lesson.

Lab13: Build and Submit Applications to YARN (Scala)

About This Lab

Objective:

Apply programming knowledge into stand-alone applications submitted to a YARN cluster

File Locations:

NA

Successful Outcome:

Build and submit a cluster-mode application to YARN

Lab Steps

Perform the following steps:

1. Build and Submit a Spark RDD application

- a. Open a terminal and use SSH to connect to sandbox:

```
# ssh sandbox
```

```
root@ubuntu:~# ssh sandbox
```

- b. **OPTIONAL:**

If you have a favorite Linux text editor already, you may use it for the rest of the lab. If you are not already familiar with a Linux text editor, we recommend that you download and install `nano` – a small, simple to use editor that will be used for the commands and screenshots in this lab.

```
yum -y install nano
```

```
[root@sandbox ~]# yum -y install nano
```

- c. Navigate to

`/root/spark/applications/scala/SparkRDD/src/main/scala/stub/` and view the `SparkRDD.scala` file.

```
# cd /root/spark/applications/scala/SparkRDD/src/main/scala/stub/
```

```
[root@sandbox ~]# cd /root/spark/applications/scala/SparkRDD/src/main/scala/stub/  
/
```

```
# nano SparkRDD.scala
```

(Again, `vi` or another editor can also be used based on your preference.)

```
[root@sandbox stub]# nano SparkRDD.scala
```

```

GNU nano 2.0.9      File: SparkRDD.scala      Modified
package stub
//ToDo
//Import the correct libraries

object SparkRDD{
  def main(args: Array[String]) {

    //Create the SparkConf
    //Set the App name to SparkRDD
    //Set the serializer to Kryo
    //Set Spark Speculation to True

    //Create the SparkContext from the conf file
    //Set LogLevel to WARN

    //Read in /user/root/selfishgiants.txt on HDFS
    //Perform wordcount
    //Print the top 10 most used words

^G Get Help   ^O WriteOut  ^R Read File  ^Y Prev Page  ^K Cut Text   ^C Cur Pos
^X Exit       ^J Justify   ^W Where Is   ^V Next Page  ^U UnCut Text ^T To Spell

```

The objective is to build an application based on this template and the comments posted on this template. You may try to do this on your own, or use the solution steps below:

Lab13: Build and Submit Applications to YARN (Scala)

```
//package stub
  package solution
//ToDo
//Import the correct libraries
  import org.apache.spark.SparkContext
  import org.apache.spark.SparkContext._
  import org.apache.spark.SparkConf

  object SparkRDD{
    def main(args: Array[String]) {
//Create the SparkConf
//Set the App name to SparkRDD

      val conf = new SparkConf().setAppName("SparkRDD")

//Set the serializer to Kryo

      conf.set("spark.serializer", "org.apache.spark.serializer.KryoSerializer")

//Set Spark Speculation to True
      conf.set("spark.speculation", "true")

//Create the SparkContext from the conf file
      val sc = new SparkContext(conf)

      /Set LogLevel to WARN
      sc.setLogLevel("WARN")

//Read in /user/root/selfishgiants.txt on HDFS
//Perform wordcount
      val input = sc.textFile("/user/root/selfishgiants.txt")
      val wc = input.flatMap(line => line.split(" ")).
        map(line => (line,1)).reduceByKey((a,b) => a+b).
        map(case (a,b) => (b,a)).sortByKey(false)
//Print the top 10 most used words
      println("You submitted the Solution")
      wc.take(10).foreach(println)
//Stop the SparkContext
      sc.stop()
    }
  }
}
```

The solution file is also available at:

/root/spark/applications/scala/SparkRDD/src/main/scala/solutions/

SolutionFileName: SparkRDD.scala (vi editor can also be used)

```
[root@sandbox ~]# cd /root/spark/applications/scala/SparkRDD/src/main/scala/solu
tion/
```

```

GNU nano 2.0.9          File: SparkRDD.scala

package solution

import org.apache.spark.SparkContext
import org.apache.spark.SparkContext._
import org.apache.spark.SparkConf

object SparkRDD{
  def main(args: Array[String]) {

    val conf = new SparkConf().setAppName("SparkRDD")
    conf.set("spark.serializer","org.apache.spark.serializer.KryoSerializer$
    conf.set("spark.sepeculation","true")

    val sc = new SparkContext(conf)
    sc.setLogLevel("WARN")

    val input = sc.textFile("/user/root/selfishgiants.txt")
    val wc = input.flatMap(line => line.split(" ")).
      map(line => (line,1)).reduceByKey((a,b) => a+b).

    Read 26 lines
  }
}
^G Get Help  ^O WriteOut  ^R Read File  ^Y Prev Page  ^K Cut Text   ^C Cur Pos
^X Exit      ^J Justify   ^W Where Is  ^V Next Page  ^U UnCut Text ^T To Spell

```

- d. Exit the text editor and save your changes (in nano, press Ctrl + X to exit and press Y to save your changes).
- e. Now to run the application first we have to package it with maven and make sure the pom.xml file has all the appropriate dependencies

```
# cd /root/spark/applications/scala/SparkRDD/
```

```
[root@sandbox ~]# cd /root/spark/applications/scala/SparkRDD/
```

```
# nano pom.xml
```

```
[root@sandbox SparkRDD]# nano pom.xml
```

```

GNU nano 2.0.9      File: pom.xml
<project xmlns="http://maven.apache.org/POM/4.0.0" xmlns:xsi="http://www.w3.org$
<modelVersion>4.0.0</modelVersion>
<groupId>com.hortonworks.SparkRDD</groupId>
<artifactId>SparkRDD</artifactId>
<version>1.0-SNAPSHOT</version>
<packaging>jar</packaging>
<build>
  <plugins>
    <plugin>
      <groupId>org.scala-tools</groupId>
      <artifactId>maven-scala-plugin</artifactId>
      <executions>
        <execution>
          <goals>
            <goal>compile</goal>
            <goal>testCompile</goal>
          </goals>
        </execution>
      </executions>
    </plugin>
  </plugins>
</build>

```

[Read 42 lines]

^G Get Help ^O WriteOut ^R Read File ^Y Prev Page ^K Cut Text ^C Cur Pos
^X Exit ^J Justify ^W Where Is ^V Next Page ^U UnCut Text ^T To Spell

Build the package:

```
# mvn package
```

A successful build will produce the following output:

```

[INFO] Building jar: /root/spark/applications/scala/SparkRDD/target/SparkRDD-1.0
-SNAPSHOT.jar
[INFO] -----
[INFO] BUILD SUCCESS
[INFO] -----

```

- f. Run the application from the terminal. Please see the tip below before running your command, as the correct command is slightly different from what is shown below.

```
spark-submit --class solution.SparkRDD --master yarn-cluster --num-executors 2
--executor-memory 1g target/SparkRDD-1.0-SNAPSHOT.jar
```

```
[root@sandbox SparkRDD]# spark-submit --class solution.SparkRDD --master yarn-cl
uster --num-executors 2 --executor-memory 1g target/SparkRDD-1.0-SNAPSHOT.jar
```



TIP:

--class packagename.objectname.

In your case this would be `stub.SparkRDD`. This would change if you ran the job from the solution file `solution.SparkRDD`.

**NOTE:**

This application will now use YARN as the resource manager with number of executors as 2 and 1g of memory.

```

ate: FINISHED)
16/06/16 10:28:09 INFO Client:
    client token: N/A
    diagnostics: N/A
    ApplicationMaster host: 172.17.0.2
    ApplicationMaster RPC port: 0
    queue: default
    start time: 1466087268180
    final status: SUCCEEDED
    tracking URL: http://sandbox:8088/proxy/application_1465503909288_0004/
    user: root
16/06/16 10:28:09 INFO Utils: Shutdown hook called
16/06/16 10:28:09 INFO Utils: Deleting directory /tmp/spark-7eff83df-d8ca-424e-83b7-99ef
a4eab46d

```

Copy the application ID at the end when the application stops.
The output of the program can be seen using the following command:

```
yarn logs -applicationId <id>
```

```
[root@sandbox SparkRDD]# yarn logs -applicationId application_1465503909288_004
```

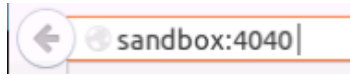
Scroll up to see the output

```

LogType:stdout
Log Upload Time:Thu Jun 16 10:28:10 -0400 2016
LogLength:110
Log Contents:
You submitted the Solution
(148,the)
(85,and)
(44,he)
(38,to)
(32,was)
(31,)
(28,in)
(22,a)
(21,were)
(19,of)
End of LogType:stdout

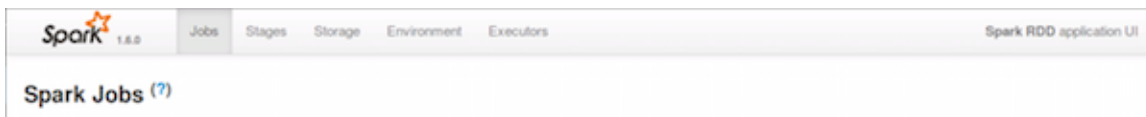
```


- g. Monitor the submitted Job. Open a new tab on the Firefox browser and browse to <http://sandbox:4040/>



IMPORTANT:

The UI below will only be available while the job is running. If you are unable to see the UI, run the application again and quickly switch to the provided link.



Result

You have successfully built and submitted a Spark application to a YARN cluster.

Lab 14: Machine Learning Walkthrough

About This Lab

Objective:

Observe and run code examples that demonstrate machine learning processes.

File Locations:

NA

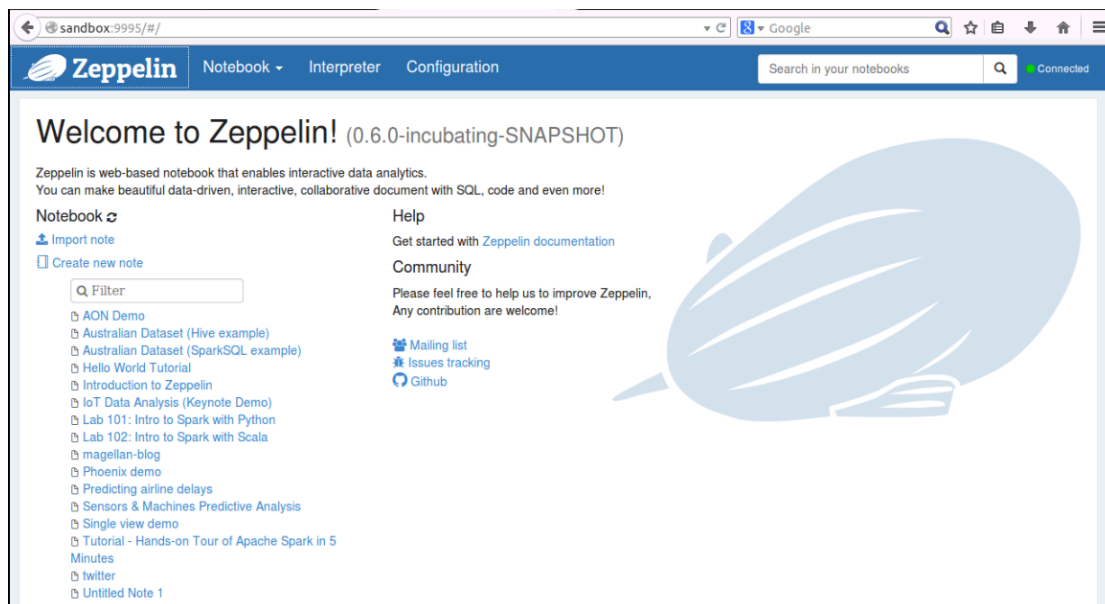
Successful Outcome:

Import a preconfigured note that contains machine learning code samples, read through the note, and run those examples.

Lab Steps

Perform the following steps:

1. Import the note, read through it, and run code examples.
 - a. Open the Firefox browser and enter the following URL to view the Zeppelin UI.
<http://sandbox:9995/>



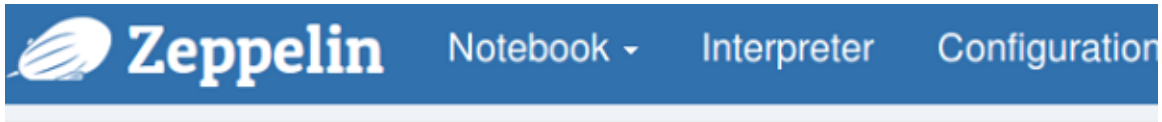
NOTE:

Zeppelin's current main backend processing engine is Apache Spark.

- b. Import a copy of the note at the following URL:

<https://raw.githubusercontent.com/hortonworks-gallery/zeppelin-notebooks/master/2BNDT63TY/note.json>

Name this note Machine Learning Lab. It should appear in the list of available notes on the Zeppelin home page.



Welcome to Zeppelin! (0.6.0-incubating)

Zeppelin is web-based notebook that enables interactive data analytics. You can make beautiful data-driven, interactive, collaborative document with SQL, code

Notebook ↻



Create new note

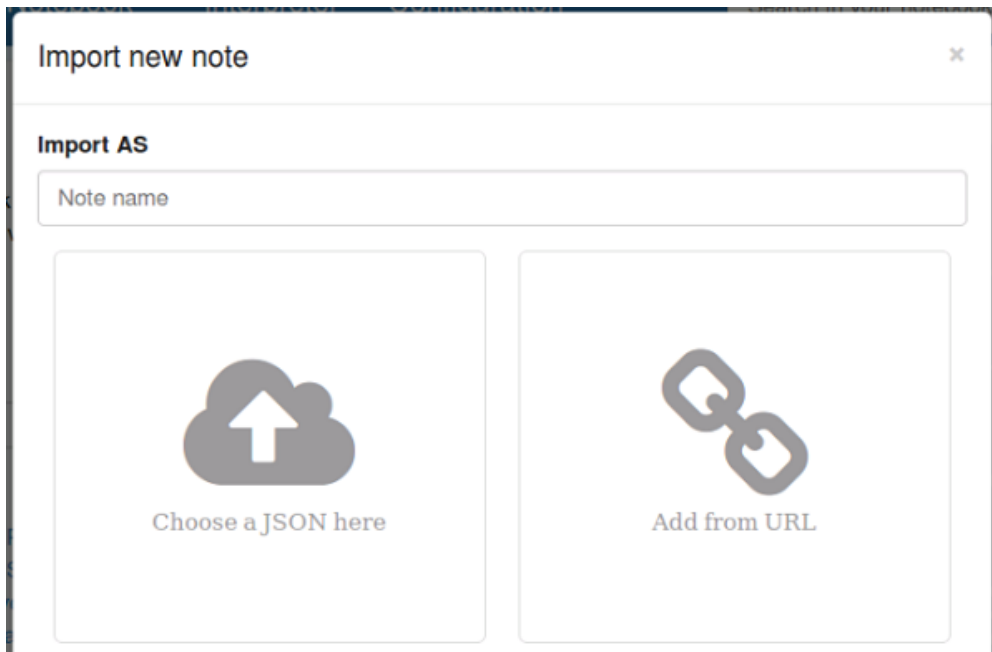
Q Filter

Help

Get started with [Zeppelin docu](#)

Community

Please feel free to help us to in



Import new note

Import AS

URL



NOTE:

If for some reason the URL is not working, your instructor should know the location of a JSON copy of this note that can be imported instead of importing it from an Internet link.

- c. Open the new note and set the interpreter to spark-yarn-client.

Zeppelin Notebook - Interpreter Configuration

Machine Learning Lab

Introduction to Machine Learning with Apache Spark

Lab created for the Hands-on guide of the [Future of Data](#) Apache Spark ecosystem

? ⚙️ 🔒 default ▾

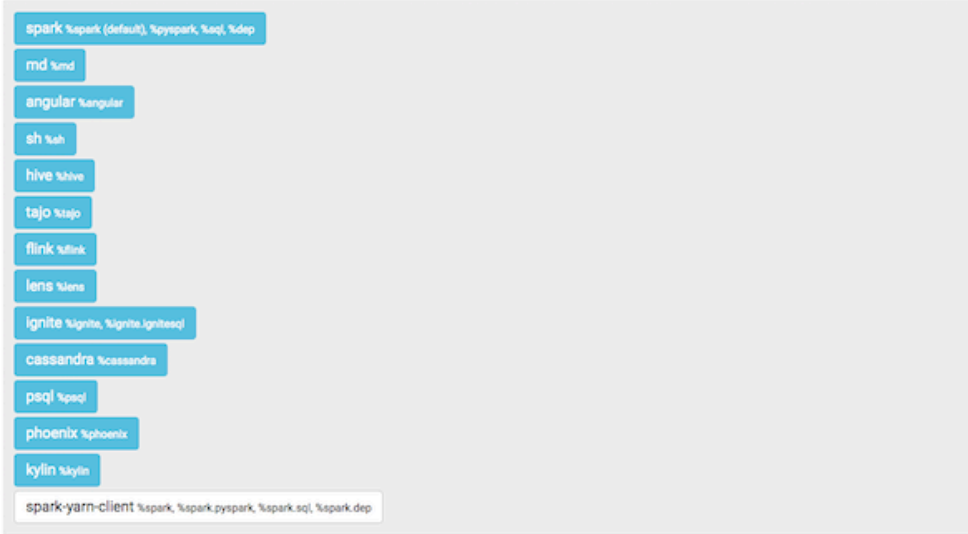
Lab 14: Machine Learning Walkthrough

Data Visualization ▶ ⌂ 🔍 📄 📁 🗑️ 🔒 🔓 default ▾

Settings

Interpreter binding

Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu.



spark %spark (default), %pyspark, %sql, %dep
md %md
angular %angular
sh %sh
hive %hive
tajo %tajo
flink %flink
lens %lens
ignite %ignite, %ignite.igniteql
cassandra %cassandra
psql %psql
phoenix %phoenix
kylin %kylin
spark-yarn-client %spark, %spark.pyspark, %spark.sql, %spark.dep

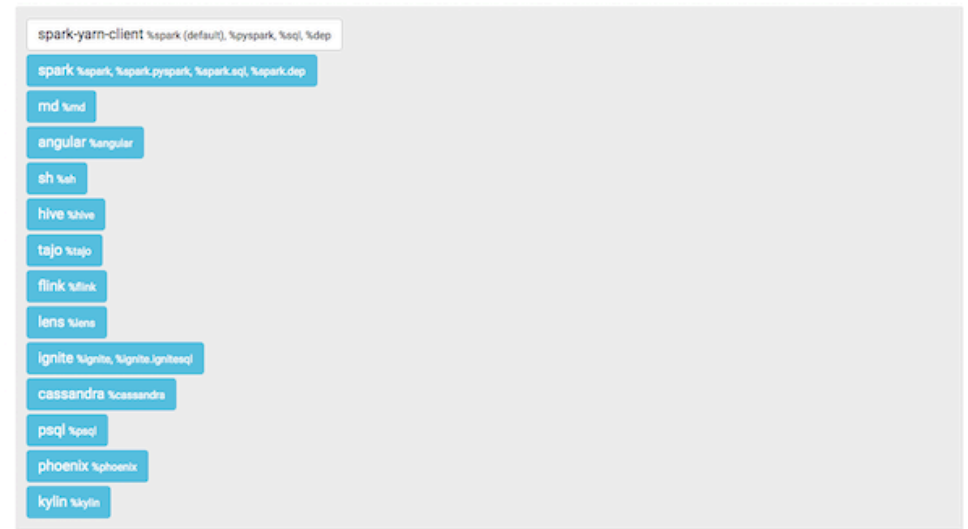
Save Cancel

Data Visualization ▶ ⌂ 🔍 📄 📁 🗑️ 🔒 🔓 default ▾

Settings

Interpreter binding

Bind interpreter for this note. Click to Bind/Unbind interpreter. Drag and drop to reorder interpreters. The first interpreter on the list becomes default. To create/remove interpreters, go to [Interpreter](#) menu.



spark-yarn-client %spark (default), %pyspark, %sql, %dep
spark %spark, %spark.pyspark, %spark.sql, %spark.dep
md %md
angular %angular
sh %sh
hive %hive
tajo %tajo
flink %flink
lens %lens
ignite %ignite, %ignite.igniteql
cassandra %cassandra
psql %psql
phoenix %phoenix
kylin %kylin

Save Cancel

Lab 14: Machine Learning Walkthrough

- d. Read through the note. A fair number of paragraphs are there for context and instructions. When you come to the first paragraph that displays code, run the code in that paragraph and view the results.

KMeans is implemented as an Estimator and generates a KMeansModel as the base model.

Note that the data points for the training are hardcoded in the example below. Before you run the K-Means sample code, try to guess what the two cluster centers should be based on the training data.

```
import org.apache.spark.ml.clustering.KMeans
import org.apache.spark.mllib.linalg.Vectors

import org.apache.spark.sql.{DataFrame, SQLContext}

val sqlContext = new SQLContext(sc)

// Creates a DataFrame
val dataset: DataFrame = sqlContext.createDataFrame(Seq(
  (1, Vectors.dense(0.0, 0.0, 0.0)),
  (2, Vectors.dense(0.1, 0.1, 0.1)),
  (3, Vectors.dense(0.2, 0.2, 0.2)),
  (4, Vectors.dense(3.0, 3.0, 3.0)),
  (5, Vectors.dense(3.1, 3.1, 3.1)),
  (6, Vectors.dense(3.2, 3.2, 3.2))
)).toDF("id", "Features")

// Trains a k-means model
val kmeans = new KMeans()
```

READY ▶ ⏏ ⏏ ⏏ ⏏

- e. Continue down the note, reading the descriptions and explanations and running the code as instructed, until you reach the end of the note.

The End

READY ▶ ⏏ ⏏ ⏏ ⏏

This concludes our lab. Hopefully you've got a taste of how easy it is to run very sophisticated clustering and classification models with Apache Spark!

Resources: Hortonworks Community Connection

READY ▶ ⏏ ⏏ ⏏ ⏏

Make sure to checkout [Hortonworks Community Connection \(HCC\)](#) if you have Apache Spark and/or Data Science / Analytics related questions or you would like to contribute back to the community with your own answers/examples/articles/repos.

All best,
The HCC Team!



Result

You have walked through a preconfigured Zeppelin note that contained multiple examples of machine learning code.



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