Microsoft

Getting Started with Microsoft R Server on HDInsight (R Server)

Contents

Introduction

Introduction

In this lab we are going demonstrate:

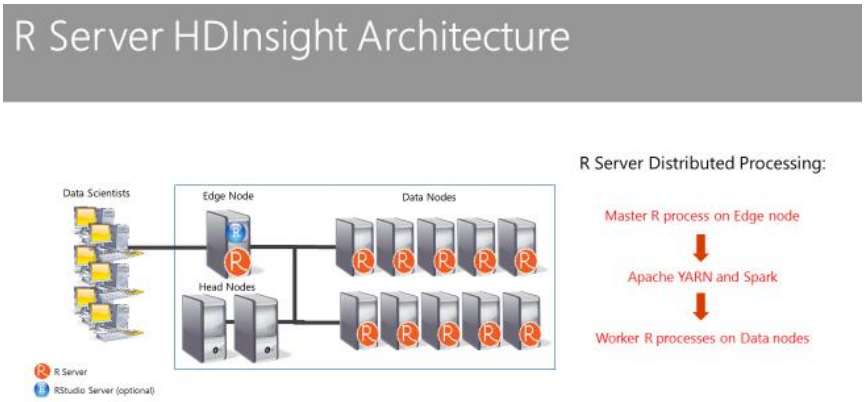
- **Checking the Yarn Memory configuration for the cluster using the cluster Dashboards.**
- **Connecting to the R Server Edge node and calling the console.**

Using R Studio (Community Edition) to run the following:

- **Running a simple test script, which articulates how MRS interacts with Hadoop i.e.**
    - **Compute on the edge node with data streamed from HDFS**
    - **Compute in the Hadoop cluster using MapReduce**
    - **Compute in the Hadoop cluster using Spark engine**
- **Getting data from CSVs into Spark DataFrames**
- **Cleaning and manipulating Spark DataFrames with SparkR**
- **Exporting Spark DataFrames to XDF**
- **Training models (logistic regression and decision forests) using ScaleR functions in Spark compute context**
- **Deploying models trained on 'big' data and operationalizing them in the cloud with Azure ML**

Prerequisites

Before starting this lab, you should have completed the accompanying documentation: *R Server on Spark - Creating a Cluster*. Having provisioned the cluster, you have the following architecture set-up in Azure:



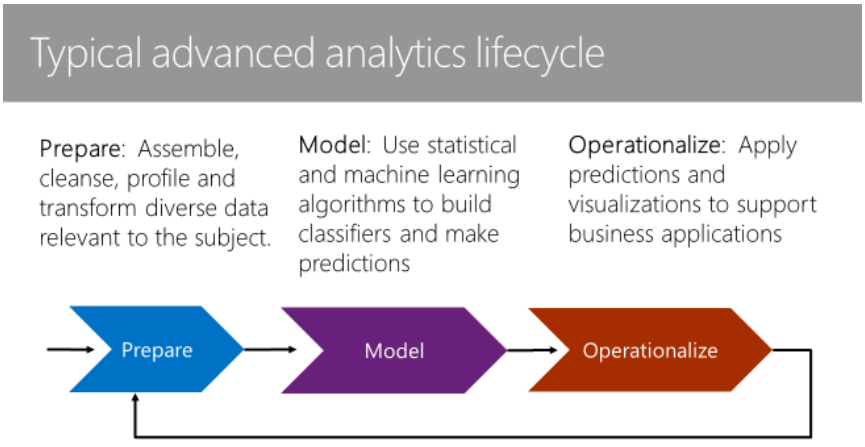
Data

We will be using

- **Airline dataset (http://www.transtats.bts.gov): contains flight information for every US commercial flight**
- **Weather dataset (http://www.ncdc.noaa.gov/orders/qclcd/): contains hourly land-based weather observations from NOAA.**

Process

We will follow a typical analytics lifecycle i.e.



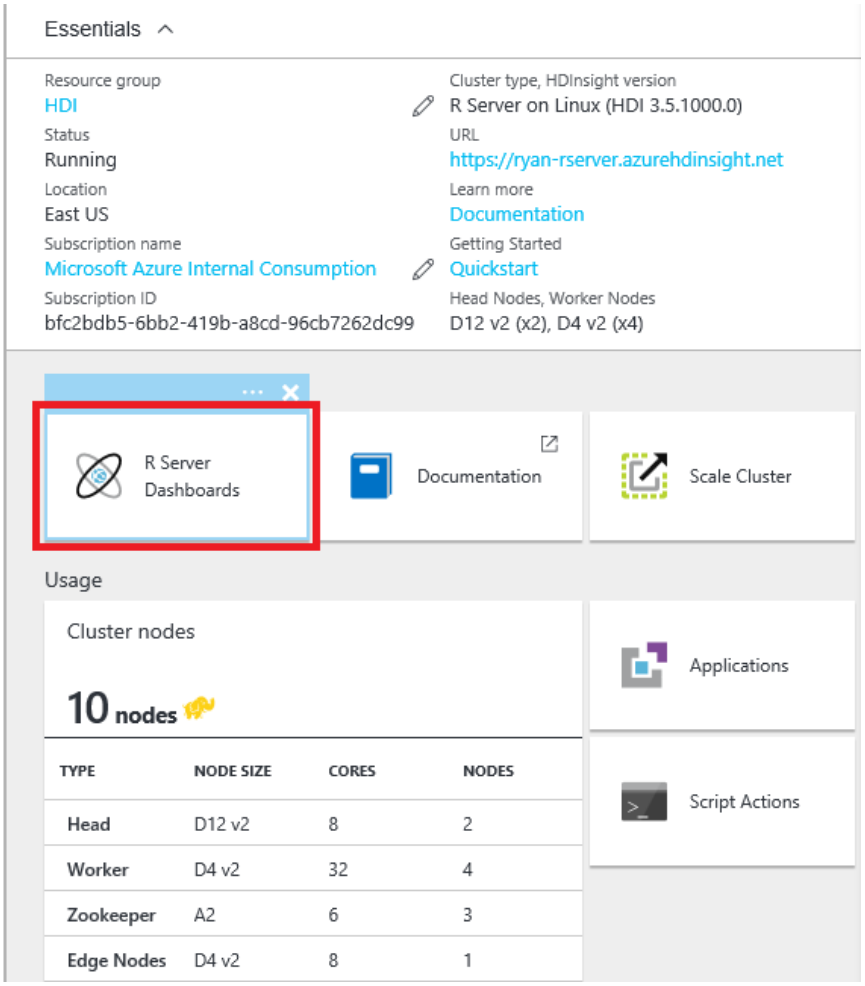
Where each step is done in a *scalable* way. For the data preparation (i.e. joining and aggregating) we use the data manipulation functions contained in the SparkR package. For the modelling we use Microsoft R Server's ScaleR package to train models using the Spark Engine. Lastly, for operationalizing we publish the model to Azure ML as a web service.

Adjust the Yarn Memory limits

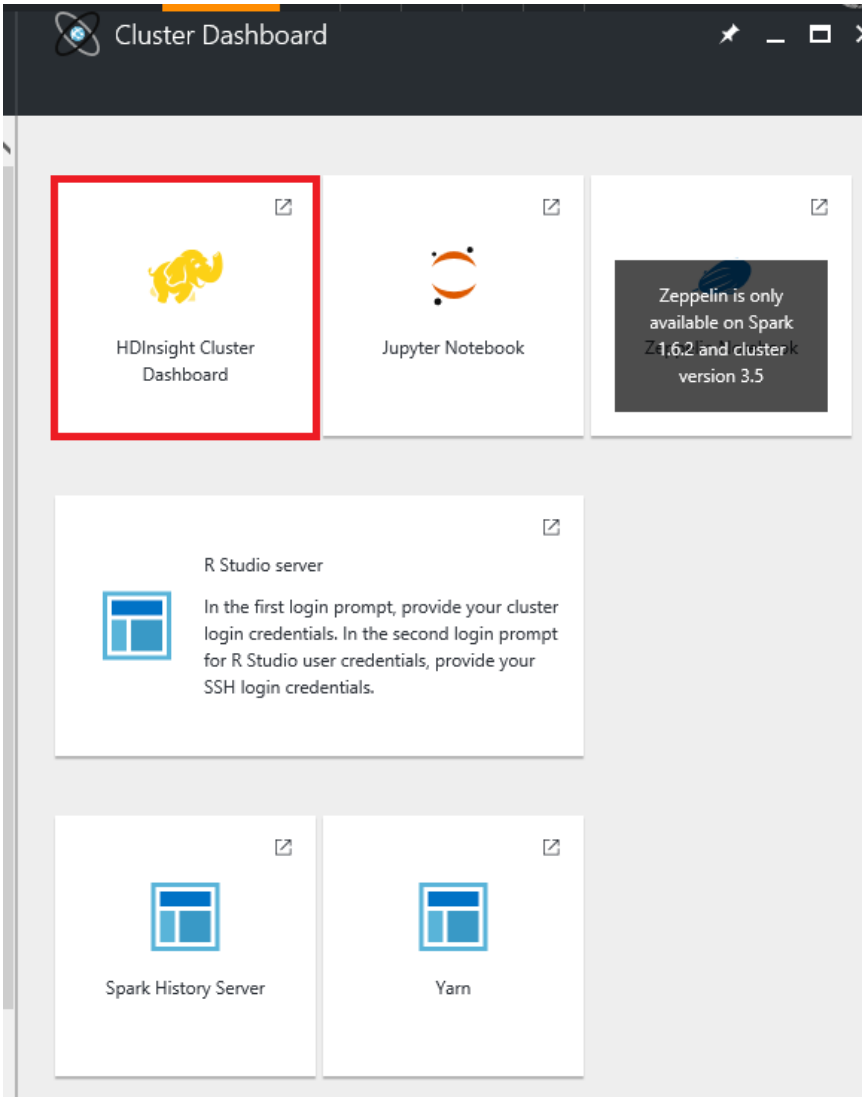
Updating the Yarn Memory configuration

In order to run the first sample script, we'll first have to ensure the memory settings for yarn.scheduler.maximum-allocation-mb' and 'yarn.nodemanager.resource.memory-mb' are at least 8192MB.

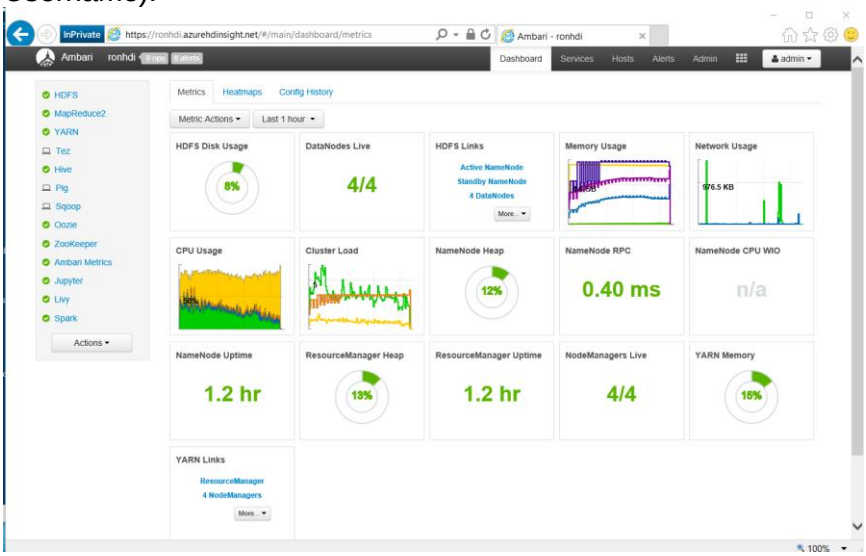
From the cluster summary page in the Azure Portal, we're going to adjust the memory settings in Ambari. You can get to ambary either from the 'Dashboard' button, or from some of the other dashboards available via 'Cluster Dashboards' -> R Server Dashboard:



Then Select 'HDInsight Cluster Dashboard' to be taken to Ambari:



You'll be logged into Ambari as the cluster admin (not the SSH Username):

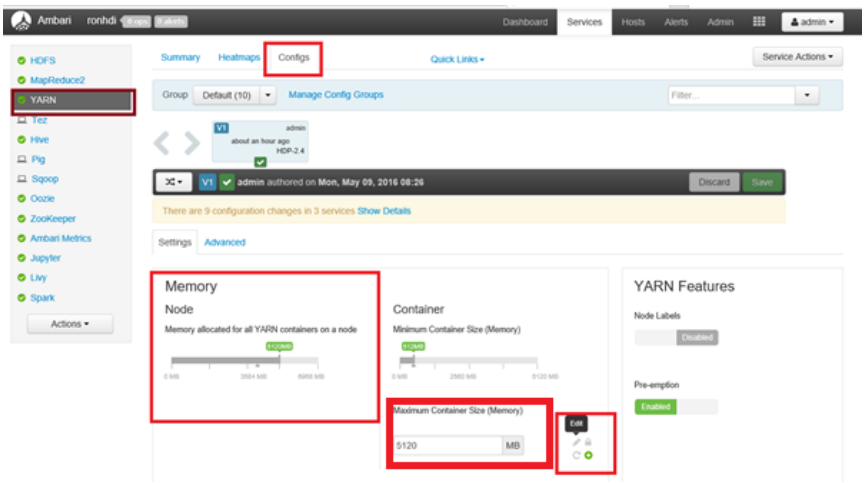


If you receive 'access denied' open a new 'in private browsing' and log back in using the Cluster Admin details.

Under 'Yarn' (left hand pane) 'Configs' tab, look at the values for Memory and Maximum Container Size.

If these settings are already greater than 9024MB, then you do not need to do the following step and you can move on to the *Connecting to the R Server Edge Node* section.

If your settings are **less than** 9024MB, then click on edit and adjust to 9024 > select **save** – proceed anyway, and select ok, restart affected, restart all and restart the services on the cluster. This may take 5 minutes or so:



Do not change any other settings using this version of Ambari, including anything regards Users, this will create unexpected behavior and may need you to recreate the cluster.

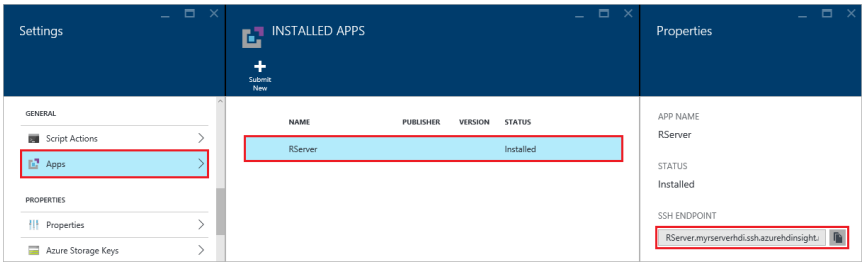
Connecting to the R Server Edge Node console

Connect to the R Server edge node

Whilst you'll run this lab using R Studio Community Edition, sometimes its useful to be able to get a Secure Shell onto the Edge node for example to create additional users, or run the R Command Line. You do not need to complete this step to use the rest of the lab, so feel free to move onto the next section if the console is not something you're interested in.

Connect to R Server edge node of the HDInsight cluster using SSH: or PuTTy (http://www.putty.org/)

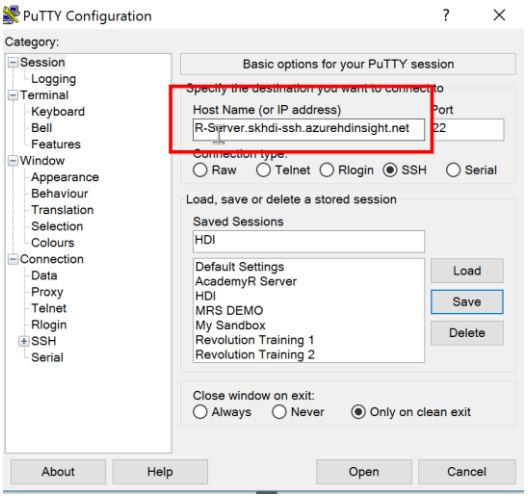
You can find the R-Server.CLUSTERNAME.ssh.azurehdinsight.net address in the Azure portal by selecting your cluster, then All Settings, Apps, and RServer. This will display the SSH Endpoint information for the edge node:



If you used a password to secure your SSH user account (not the cluster admin), you will be prompted to enter it. If you used a public key, you may have to use the -i parameter to specify the matching private key. For example,

```
ssh -i ~/.ssh/id_rsa USERNAME@RServer.CLUSTERNAME-ssh.azurehdinsight.net.
```

or in PuTTY you would enter the SSH endpoint information in the hostname text field:



For more information on using SSH with Linux-based HDInsight, see the following articles:

•Use SSH with Linux-based Hadoop on HDInsight from Linux, Unix, or OS X

•Use SSH with Linux-based Hadoop on HDInsight from Windows

Test the R Console

For this lab we'll be following R Studio, however, testing the R console is a quick way to ensure your edge node is up and running. From the SSH session, use the following command to start the R console:

```
R
```

You will see output similar to the following:

```
R version 3.2.2 (2015-08-14) -- "Fire Safety"
Copyright (C) 2015 The R Foundation for Statistical Computing
Platform: x86_64-pc-linux-gnu (64-bit)

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

Natural language support but running in an English locale

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

Microsoft R Server version 8.0: an enhanced distribution of R
Microsoft packages Copyright (C) 2016 Microsoft Corporation

Type 'readme()' for release notes.

>
```

From the > prompt, you can enter R code. R server includes packages that allow you to easily interact with Hadoop and run distributed computations. For example, use the following command to view the root of the default file system for the HDInsight cluster.

```
rxHadoopListFiles("/")
```

Note that in this instance, the HDFS store is actually Azure Blob store, so you can upload files to the cluster via blob store and process them with R at scale!

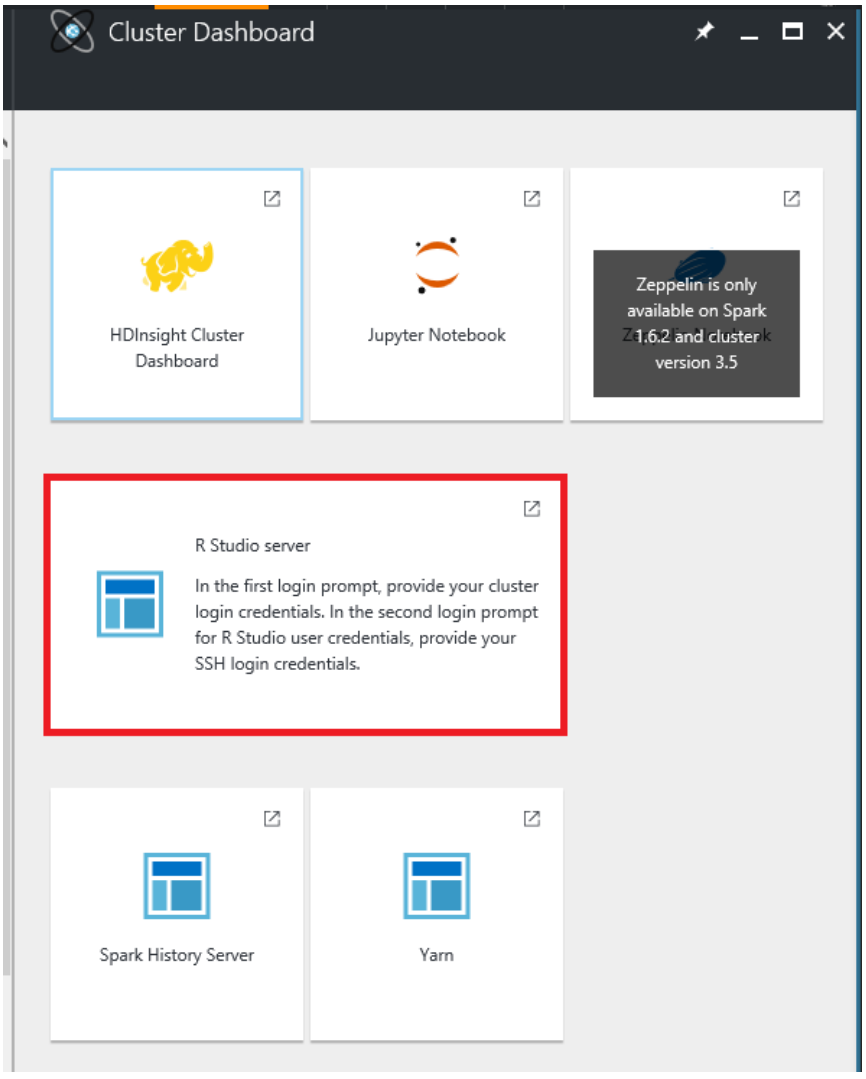
Working with R Studio Community Edition

Overview

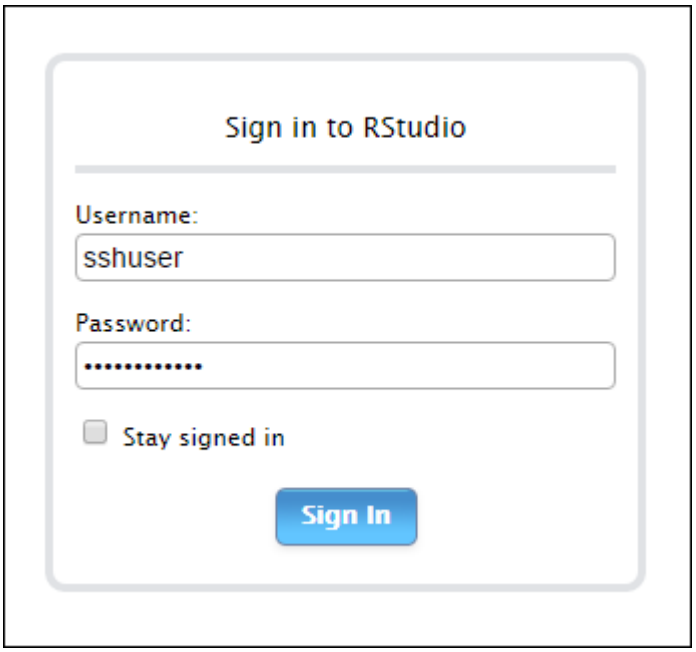
There are multiple integrated development environments (IDE) available for R today, including Microsoft's recently announced R Tools for Visual Studio (RTVS), a family of desktop and server tools from RStudio, or Walware's Eclipse-based StatET. Among the most popular on Linux is the use of RStudio Server that provides a browser-based IDE for use by remote clients. HDInsight installs RStudio Server community edition on the edge node of an HDInsight Standard R Server cluster which provides a full IDE experience for the development and execution of R scripts with R Server on the cluster, and can be considerably more productive than default use of the R Console. Note that the community edition allows only 1 user at a time to be interactive with the console, if a second user tries to connect then the previous logged in will be logged off.

Opening RStudio

1. You can browse to https://<your-cluster-name>.azurehdinsight.net/rstudio or from the cluster dashboard pane you can select 'R Studio Server'

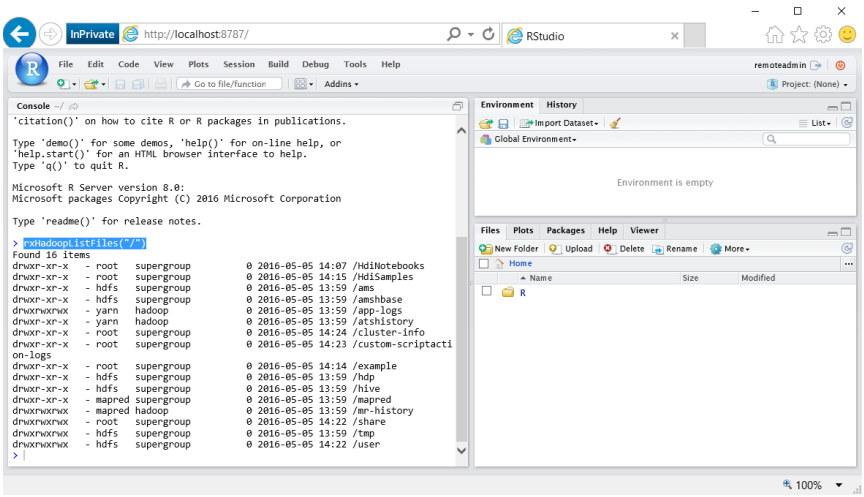


1. You'll be prompted twice to log in – first for plain text authentication to authenticate with the cluster (use the local **admin cluster login**), then second to actually sign into the studio, use the **SSH user**.

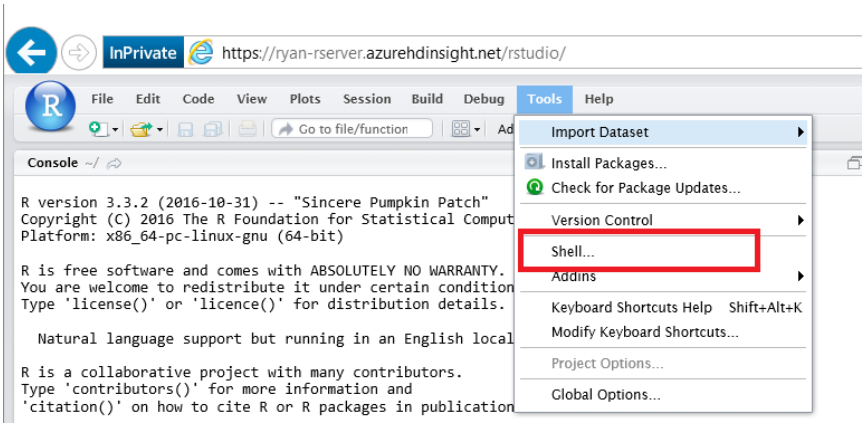


2. As with the command prompt, You can browse the HDFS file system using

```
rxHadoopListFiles("/");
```



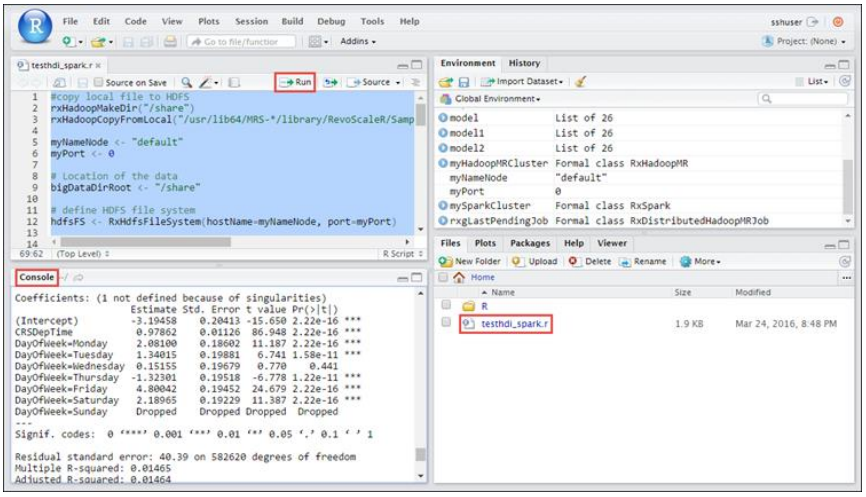
To test whether the RStudio installation was successful, you can run a test script that executes R based MapReduce and Spark jobs on the cluster. You can run the following from a remote shell session, or use the shell tool in R Studio by selecting ‘Tools.. shell’



3. Run the following from your shell to download the test script:

```
wget
http://mrsactionscripts.blob.core.windows.net/rstudio
-server-community-v01/testhdi_spark.r
```

4. In RStudio, you will see the test script **(testhdi_spark.r)** you downloaded. Double click the file to open it, select the contents of the file using ‘ctrl-a’, and then click Run. You should see the output in the Console pane.



This test script shows running in **"local" mode and streaming data from HDFS** i.e.

```
# define the data source
airDS <- RxTextData(file = inputFile, missingValueString = "M",
                    colInfo = colInfo, fileSystem = hdfsFS)

# First test the "local" compute context
rxSetComputeContext("local")

# Run a linear regression
system.time(
  model <- rxLinMod(ArrDelay~CRSDepTime+DayOfWeek, data = airDS)
)
```

The test script shows running the same linear model as a **MapReduce** job, by just switching the compute context i.e.

```
# define MapReduce compute context
myHadoopMRCluster <- RxHadoopMR(consoleOutput=TRUE,
                                nameNode=myNameNode,
                                port=myPort,
                                hadoopSwitches="-libjars /etc/hadoop/conf")

# set compute context
rxSetComputeContext(myHadoopMRCluster)

# Run a linear regression
system.time(
  model1 <- rxLinMod(ArrDelay~CRSDepTime+DayOfWeek, data = airDS)
)
```

The test script also demonstrates running the same linear model as a **Spark job** i.e.

```
# define Spark compute context
mySparkCluster <- RxSpark(consoleOutput=TRUE)

# set compute context
rxSetComputeContext(mySparkCluster)

# Run a linear regression
system.time(
  model2 <- rxLinMod(ArrDelay~CRSDepTime+DayOfWeek, data = airDS)
)
```

Once you have finished with this script, switch back to local compute context by issuing the following command:

```
> rxSetComputeContext("localpar")
```

Downloading the Sample Data and loading into HDFS

Overview

There is some sample airline data for you to clean, join and then predict on using R Studio, along with the required script files.

Before continuing you'll need to load them onto the R Server Edge Node.

Loading the files

From the Shell Tool (either Putty or Tools.. shell)

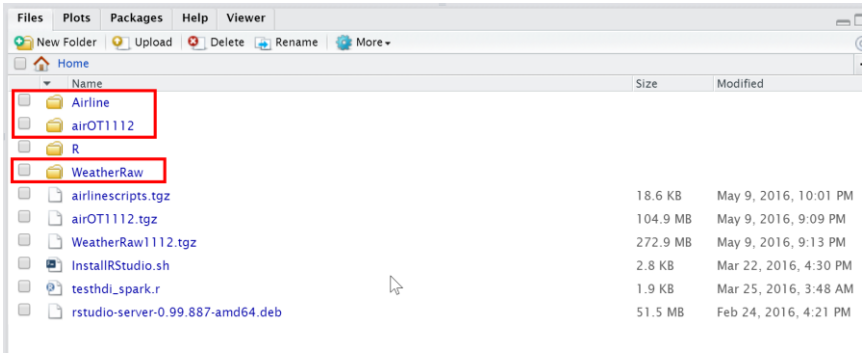
Run the following:

```
wget
https://msrlabdemos.blob.core.windows.net/airlinesubset/airOT1112.tgz

wget
https://msrlabdemos.blob.core.windows.net/airlinesubset/WeatherRaw1112.tgz

wget
https://msrlabdemos.blob.core.windows.net/airlinesubset/airlinescripts.tgz
```

Once downloaded, unpack on the R node to the home directory (default) using the following commands:

```
tar -xvf WeatherRaw1112.tgz
tar -xvf airOT1112.tgz
tar -xvf airlinescripts.tgz
```

If you haven't already done so, log into R Studio. You should see the unzipped folders in your home directory:



At the R Console set your working directory to the Airline directory, i.e.

```
Console ~/Airline/
> setwd("~/Airline")
```

In the **Airline** directory you will see an R script called **LoadData.r** > Open and run this script line-by-line in R. Below is a description of what the LoadData.r script is doing:

List the files in HDFS – do they exist?

Make a directory called WeatherRaw in HDFS and copy the local (linux) files to that location

Do the same for the airlines data

```
# understand where the data is in the HDFS store:
rxHadoopListFiles("/user/RevoShare/WeatherRaw")
rxHadoopListFiles("/share")

#use getwd() to establish what the current working directory is. use setwd if necessary

rxHadoopMakeDir("/user/RevoShare/WeatherRaw")
rxHadoopCopyFromLocal("../WeatherRaw/*", "/user/RevoShare/WeatherRaw/")
rxHadoopListFiles("/user/RevoShare/WeatherRaw")

rxHadoopMakeDir("/user/RevoShare/AirOnTime08to12CSV")
rxHadoopCopyFromLocal("../airOT1112/*", "/user/RevoShare/AirOnTime08to12CSV/")
rxHadoopListFiles("/user/RevoShare/AirOnTime08to12CSV")
```

The data files now exist in HDFS and this means that we can use them in Spark and/or MapReduce.

Running the scripts

The Scripts can be found in the ~/Airline Directory:

1. Run script **0-Weather-Preprocess.r** (do this 'command-by-command' to get an understanding of what is running). This script uses an rxDataStep (in a spark Compute context) to adjust time in the airline dataset to be the same as the weather dataset.

2. Run script **1-Clean-Join.r** (do this 'command-by-command' to get an understanding of what is running). In this script we are using SparkR is aggregate and join datasets. Once we are happy with the dataset, we export it to XDF for training a predictive model.

   Exercises:

   a. For the SparkR DataFrame airDF, can you create a new DataFrame containing only observations from the year 2012

      Hint: the SparkR package has a filter function. More details can be found on the help page here.

   b. Can you compute (using SparkR) the average Arrival Delay (ARR_DELAY column) by year (YEAR column)?

      Hint: Look up the groupBy function in the documentation here for an example. You should have something like:

      mydf <- agg(groupBy(),…)

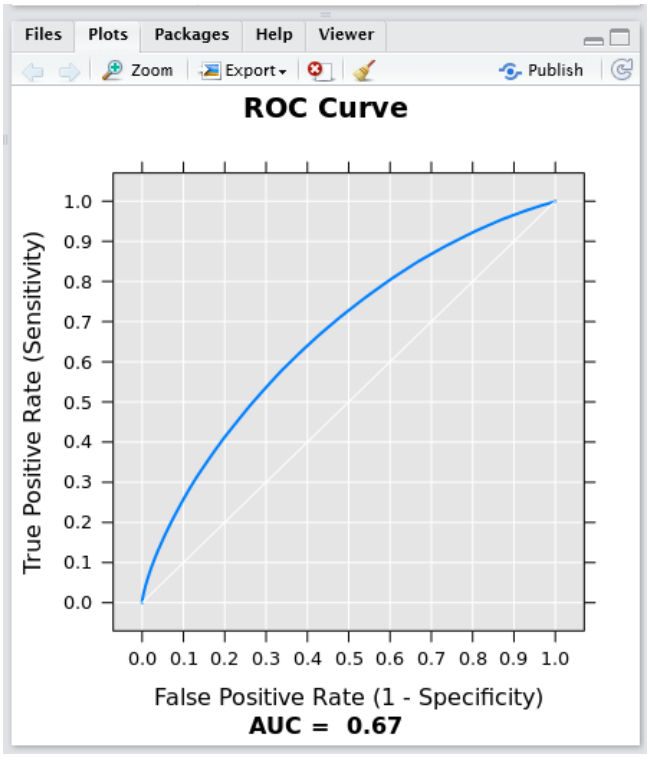
      When you do your aggregation does anything run in Spark? What does mydf look like?

      How we get Spark to run the aggregation and pull back to a native data.frame?

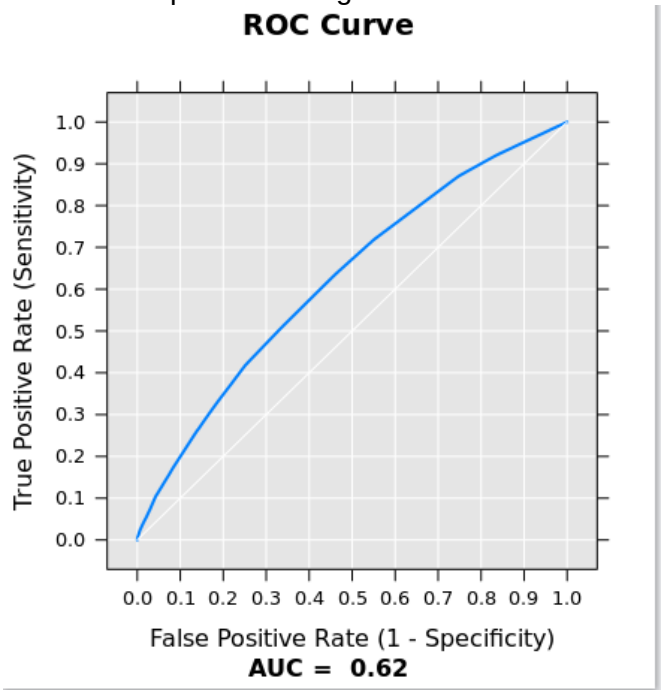
      Hint: Look up "Under the Hood: Intelligent Optimization and Code Generation" here. A DataFrame can be coerced to a data.frame in R using as.data.frame(). In this case the aggregated data is small, but we must be careful with other cases where the data could be huge and we risk blowing RAM on the edge node.

3. Run script **2-Train-Test.r** (do this 'command-by-command' to get an understanding of what is running). In this script we split our XDF data into training and test sets. We build a logistic regression and decision tree (in Spark) on the training set. We use rxPredict and rxRoc to test and validate the model with Receiver Operator Curves.

   i.e. you will see the following plots when the ROCs are plotted



Output for the regression model



The plot for the decision trees – note the slightly lower area under the curve

Exercises:

a. Can you create a wrapper function to predict to send to Azure ML?

   Hint: you will need to coerce the dTreeModel object to an open source equivalent because Azure ML does not yet use ScaleR functions. To coerce, use:

```
rpartModel <- as.rpart( dTreeModel )
```

b. Publish the wrapper function to AzureML as a web service.

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