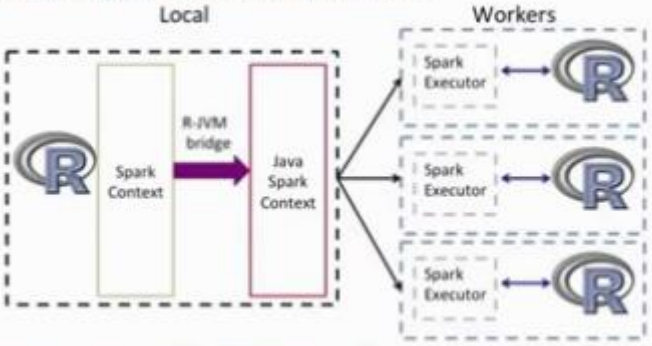
Lesson objectives

- Learn what SparkR is
- Learn why you would use SparkR
- List the features of SparkR
- Understand the interfaces into SparkR

What is SparkR?

- An R package to use Apache Spark from R



- SparkR implements distributed dataframes
- SparkR supports operations like selection, filtering, aggregation etc.
- SparkR also supports MLIB

Why use SparkR?

- R has usability issues with big data workflows
- R is very resource constrained so limited optimizations
- R has restricted machine learning

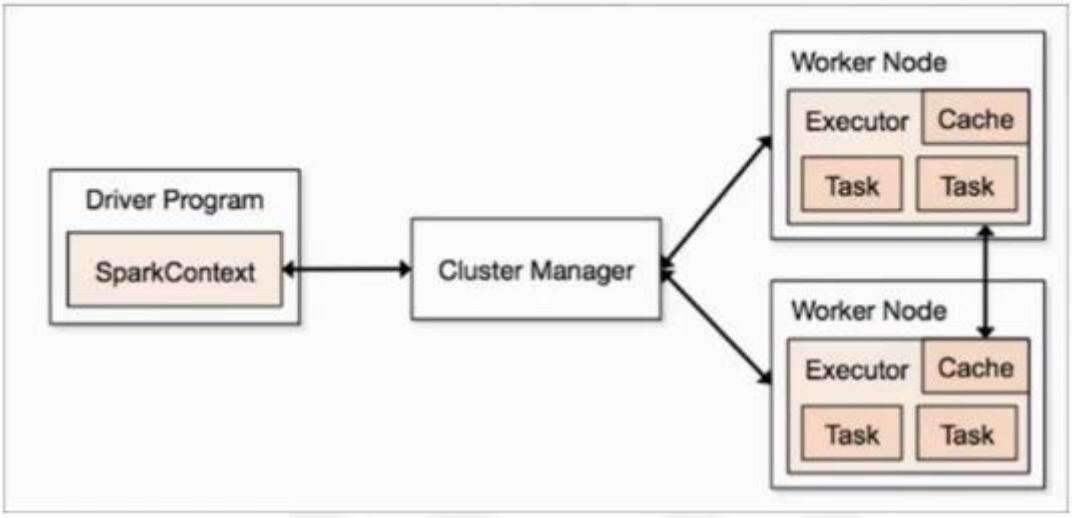
Features of SparkR

- Scalability to many cores and machines
- Data Frame Optimizations
- Data Sources API
- RDDs as Distributed Lists
- Serializing closures
- Using existing R packages

Interfaces

- Spark shell
- SparkR shell
- Rstudio
- Notebooks
- Data Scientist Workbench (an interactive service providing notebooks interface)

SparkContext



Spark SQL

- Spark SQL is a Spark module for structured data processing
- The entry point into all relational functionality in Spark is the SQLContext
- There are several ways to interact with Spark SQL including SQL, the DataFrames API and the Datasets API:
  - SQL
  - JDBC/ODBC
  - Datasets with a strongly-typed LINQ-like Query DSL
- SparkR works solely with DataFrames which requires Spark SQL

Spark shell

- The following command is used to open Spark shell:
  - Spark-shell

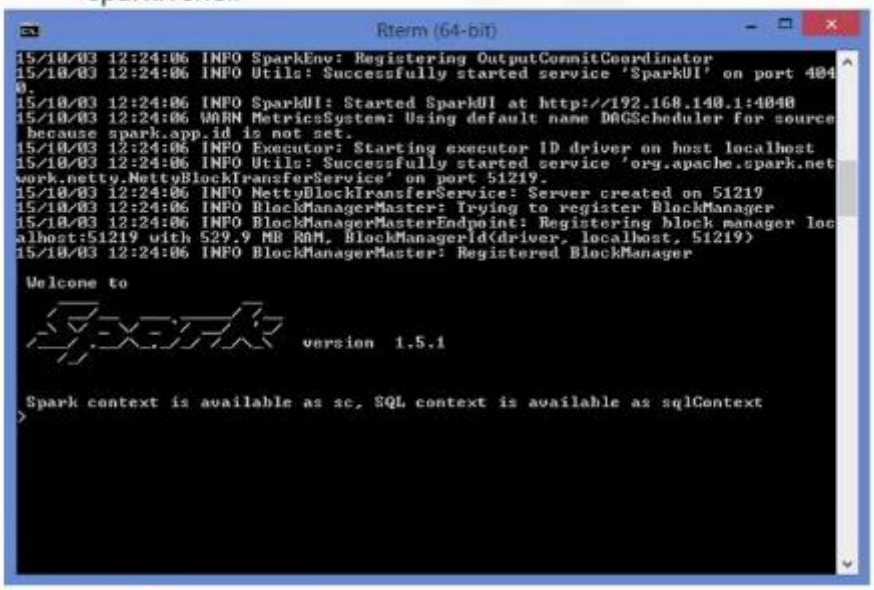
```
$ ./bin/spark-shell
Spark context available as sc.
SQL context available as sqlContext.
Welcome to
      ____              __
     / __/__  ___ _____/ /__
    _\ \/ _ \/ _ `/ __/  '_/
   /___/ .__/\_,_/_/ /_/\_\   version 1.6.0-SNAPSHOT
      /_/

Using Scala version 2.11.7 (Java HotSpot(TM) 64-Bit Server VM, Java 1.8.0_66)
Type in expressions to have them evaluated.
Type :help for more information.

scala>
```

- create a SparkContext and an sqlContext

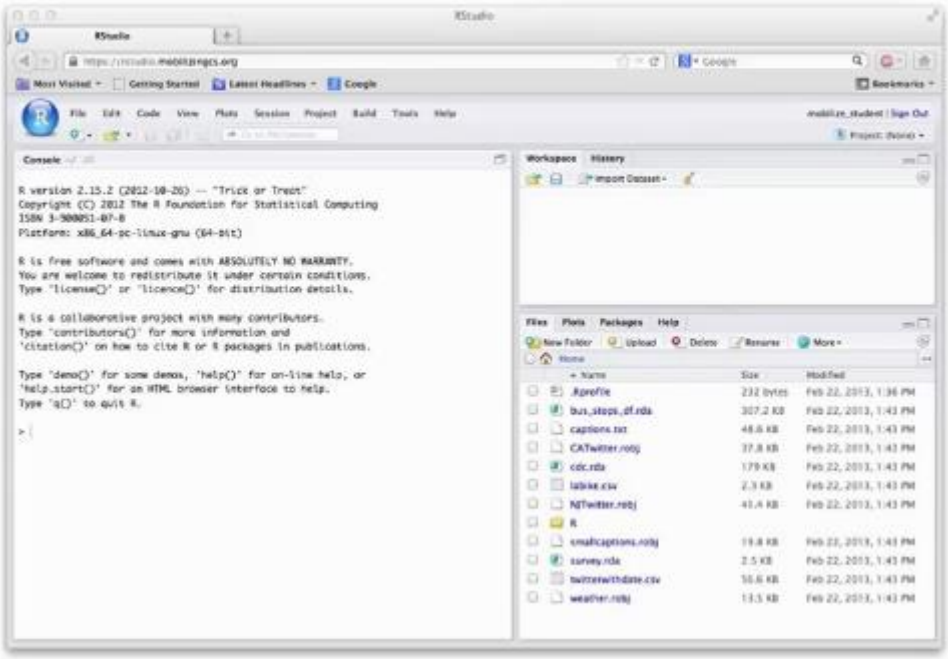
```
sc <- sparkR.init()
sqlContext <- sparkRSQL.init(sc)
```

SparkR shell

- The following command is used to open SparkR shell:
  - sparkR shell

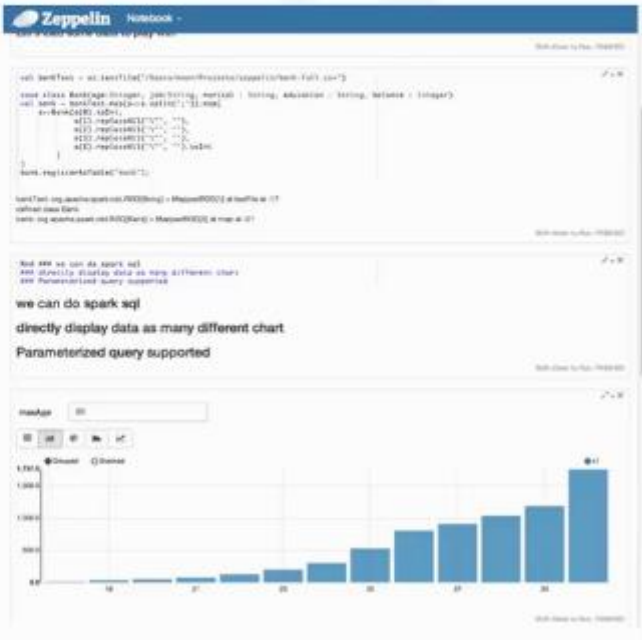


RStudio

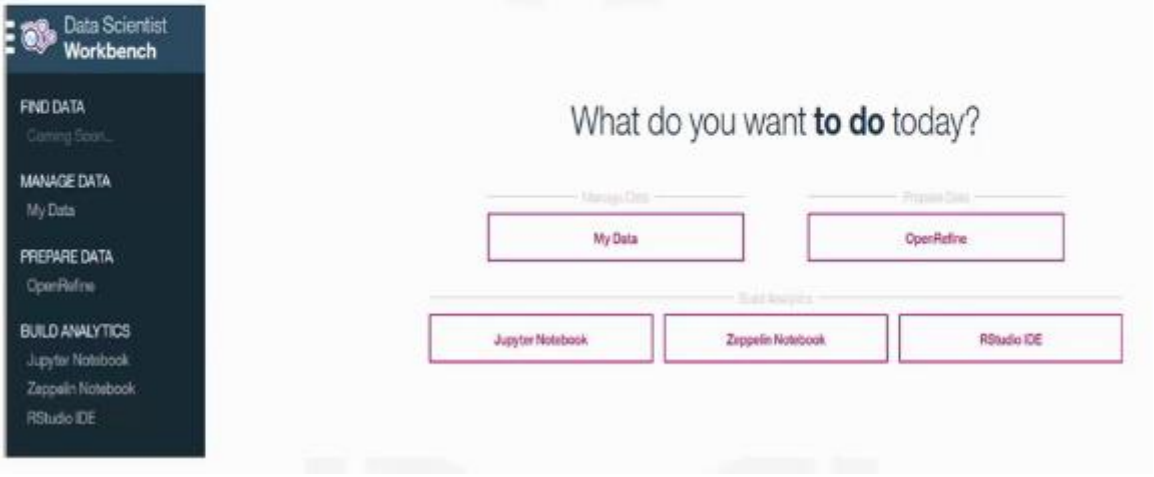


Notebooks

- Zeppelin notebook



Data Scientist Workbench (DSWB)



Lesson objectives

- Understand how to use dataframes
- Learn to select data
- Learn to filter data
- Learn to aggregate data
- Learn to operate on columns
- Understand how to write SQL queries

Dataframes

- Why dataframes
- Creating dataframes
  - Local dataframes
  - Dataframes from raw data
  - Dataframes from data sources

Name Conflicts

- It is possible to have a name conflict between functions in SparkR and R
- The following have a conflict:
  - cov in package:stats
  - filter in package:stats
  - sample in package:base
  - table in package:base

- It is also possible to have name conflicts between dplyr and SparkR

Select columns

- select

SparkR::head(select(cars,cars$mpg))

```
Out[3]:    mpg
         1 21.0
         2 21.0
         3 22.8
         4 21.4
         5 18.7
         6 18.1
```

- selectExpr

SparkR:: head(selectExpr(df, "Day", "Day + 99", "Day * 3"))

```
  Day (Day + 99) (Day * 3)
1   1        100         3
2   2        101         6
3   3        102         9
4   4        103        12
5   5        104        15
6   6        105        18
```


Operating on columns

- Operating on a specific column using $

cars$mpg <- cars$mpg/3.78541178

```
Out[4]:        mpg cyl disp  hp drat    wt  qsec vs am gear carb
         1 5.547613   6  160 110 3.90 2.620 16.46  0  1    4    4
         2 5.547613   6  160 110 3.90 2.875 17.02  0  1    4    4
         3 6.023123   4  108  93 3.85 2.320 18.61  1  1    4    1
         4 5.653282   6  258 110 3.08 3.215 19.44  1  0    3    1
         5 4.940017   8  360 175 3.15 3.440 17.02  0  0    3    2
         6 4.781514   6  225 105 2.76 3.460 20.22  1  0    3    1
```

- Operating on multiple columns using [ ]

[, c("you" , "me")]

Filter (select rows)

- filter by condition for one column

  filtered <- SparkR::head(SparkR::filter(cars, df$cyl == 6)
  collect(filtered)

```
   mpg cyl  disp  hp drat    wt  qsec vs am gear carb
1 21.0   6 160.0 110 3.90 2.620 16.46  0  1    4    4
2 21.0   6 160.0 110 3.90 2.875 17.02  0  1    4    4
3 21.4   6 258.0 110 3.08 3.215 19.44  1  0    3    1
4 18.1   6 225.0 105 2.76 3.460 20.22  1  0    3    1
```

- filter by condition for multiple columns
  - These two expression are same:

    SparkR::head(SparkR:: filter("cars, df$cyl == 6 and cars, cars$mpg < 20")

    SparkR::head(SparkR:: filter(cars, df$cyl == 6). SparkR::filter(cars, cars$mpg < 20)

SparkSQL (row operations)

- Dataframes as SQL tables

  registerTempTable(cars,"cars")
  SparkR::head(sql(sqlContext, "SELECT * FROM cars WHERE cyl > 6"))

```
   mpg cyl  disp  hp drat   wt  qsec vs am gear carb
1 18.7   8 360.0 175 3.15 3.44 17.02  0  0    3    2
2 14.3   8 360.0 245 3.21 3.57 15.84  0  0    3    4
3 16.4   8 275.8 180 3.07 4.07 17.40  0  0    3    3
```

Aggregate data

- groupBy for one column (count, sum, average)
  - groupBy(cars, cars$mpg), count = n(cars$mpg))
  - groupBy(cars, cars$mpg), sum = sum(cars$mpg))
  - groupBy(cars, cars$mpg), average = avg(cars$hp))

- groupBy for multiple columns (count, sum, average)
  - groupBy(cars, "'class', 'year'"), avg = n('hwy'))

Lesson objectives

- Understand machine learning
- Learn how to use GLM model

Machine learning

- Example machine learning workflow

- Observation = rows in a dataframe
- Feature = columns in a dataframe

MLlib and ML

- MLlib is Spark's machine learning (ML) library
- It divides into two packages:
  - spark.mllib contains the original API built on top of RDDs.
  - spark.ml provides higher-level API built on top of DataFrames
- Concepts in pipelines:
  - DataFrame
  - Transformer
  - Estimator
  - Pipeline
  - Parameter

Pipelines components

- Transformers
  - A Transformer is an abstraction that includes feature transformers and learned models
- Estimators
  - An Estimator abstracts the concept of a learning algorithm that trains on data
- Pipeline
  - a sequence of algorithms to process and learn from data

Implementing Linear Models in SparkR

- Prepare and load data
  - Load the data
  - Read the data in a Spark dataframe
  - Create factors
- Train the model
  - Formula
  - Dataset
  - Model
- Evaluate the model
  - Baseline reference
  - Predict()
- Implement model
  - Iterate through the training dataset
  - Find acceptable model

GLM data load

- Download and save the data files using R
  - population_data_files_url <- 'http://www2.census.gov/acs2013_1yr/pums/csv_pus.zip'
  - library(RCurl)
  - population_data_file <- getBinaryURL(population_data_files_url)
  - population_data_file_path <- '/nfs/data/2013-acs/csv_pus.zip'
  - population_data_file_local <- file(population_data_file_path, open = "wb")
- sparkContext
  - sc <- sparkR.init()
- sqlContext
  - sqlContext <- sparkRSQL.init(sc)

GLM data prep

- Convert any categorical variable from a numeric variable into a factor for example
  - housing_df$ST <- cast(housing_df$ST, "string")
- Sanitize the data for example remove nulls
  - housing_with_valp_df <- filter(housing_df,
    isNotNull(housing_df$VALP)
    & isNotNull(housing_df$TAXP)
    & isNotNull(housing_df$INSP)
    & isNotNull(housing_df$ACR))
- Create training and test data
  - housing_df_test <- sample(housing_with_valp_df,FALSE,0.1)

GLM model training

- glm()
  - Fit a gaussian GLM model over the dataset.
    - model <- glm(Sepal_Length ~ Sepal_Width + Species, data = df, family = "gaussian")
- summary()
  - Model summary are returned in a similar format to R's native glm().
    - summary(model)
    - ##$devianceResiduals
    - ## Min Max
    - ## -1.307112 1.412532
    - etc.....

GLM model evaluation

- predict()
  - > preds <- predict(model, training)
  - > errors <- select(
  - preds, preds$label, preds$prediction, preds$aircraft_type,
  - alias(preds$label - preds$prediction, "error"))

GLM model vizualization

- matplotlib
- Ggplot2
- Third party libraries such as prettyplotlib