About Apache Spark and Databricks

**Apache Spark** is a general distributed computing engine that unifies:

- Real-time streaming (Spark Streaming)
- Machine learning (SparkML/MLLib)
- SQL (SparkSQL)
- Graph processing (GraphX)

**Databricks Inc.** is a company founded by creators of Spark focused on making big data simple by offering an end to end data processing platform in the cloud.
What is R?

Language and runtime

The cornerstone of R is the data frame concept
Many data scientists love R

- Open source
- Highly dynamic
- Interactive environment
- Rich ecosystem of packages
- Powerful visualization infrastructure
- Data frames make data manipulation convenient
- Taught by many schools to stats and computing students
Performance Limitations of R

R language

- R’s dynamic design imposes restrictions on optimization

R runtime

- Single threaded
- Everything has to fit in memory
What would be ideal?

**Seamless manipulation and analysis of very large data in R**

- R’s flexible syntax
- R’s rich package ecosystem
- R’s interactive environment
- Scalability (scale up and out)
- Integration with distributed data sources / storage
Augmenting R with other frameworks

In practice data scientists use R in conjunction with other frameworks (Hadoop MR, Hive, Pig, Relational Databases, etc)

1. Load, clean, transform, aggregate, sample
2. Save to local storage
3. Read and analyze in R

Iterate
What is SparkR?

An R package distributed with Apache Spark:

• Provides R frontend to Spark
• Exposes Spark Dataframes (inspired by R and Pandas)
• Convenient interoperability between R and Spark DataFrames
How does SparkR solve our problems?

1. Load, clean, transform, aggregate, sample
2. Save to local storage
3. Read and analyze in R

- No local storage involved
- Write everything in R
- Use Spark’s distributed cache for interactive/iterative analysis at speed of thought
Example SparkR program

# Loading distributed data
df <- read.df("hdfs://bigdata/logs", source = "json")

# Distributed filtering and aggregation
errors <- subset(df, df$type == "error")
counts <- agg(groupBy(errors, df$code), num = count(df$code))

# Collecting and plotting small data
qplot(code, num, data = collect(counts), geom = "bar", stat = "identity") + coord_flip()
Overview of SparkR API

**IO**
- `read.df` / `write.df`
- `createDataFrame`

**Caching**
- `cache` / `persist` / ` unpersist`
- `cacheTable` / `uncacheTable`

**Utility functions**
- `dim` / `head` / `take`
- `names` / `rand` / `sample`

**ML Lib**
- `glm` / `predict`

**DataFrame API**
- `select` / `subset` / `groupBy`
- `head` / `collect` / `showDF`
- `unionAll` / `agg` / `avg` / `column`

**SQL**
- `sql` / `table` / `saveAsTable`
- `registerTempTable` / `tables`
SparkR architecture

Spark Driver

JVM

R Backend

Worker

JVM

Worker

JVM

Data Sources

cassandra

Hadoop

HBase

S3
Moving data between R and JVM

- \texttt{SparkR::collect()} from R to JVM
- \texttt{SparkR::createDataFrame()} from JVM to R Backend
Moving data between R and JVM
### Moving between languages (notebooks)

#### R

```r
df <- read.df(...)  
wiki <- filter(df, ...)  
registerTempTable(wiki, "wiki")
```

#### Scala

```scala
val wiki = table("wiki")

val parsed = recent.map {
  Row(_, _, text: String, _, _) => text.split(‘ ’)
}

val model = Kmeans.train(parsed)
```
Example use case: exploratory analysis

- Data pipeline implemented in Scala/Python
- New files are appended to existing data partitioned by time
- Table scheme is saved in Hive metastore
- Data scientists use SparkR to analyze and visualize data
  1. `refreshTable(sqlContext, "logsTable")`
  2. `logs <- table(sqlContext, "logsTable")`
  3. Iteratively analyze/aggregate/visualize using Spark & R DataFrames
  4. Publish/share results
Demo
Summary

1. SparkR is an R frontend to Apache Spark
2. Distributed data resides in the JVM
3. Workers are not running R process (yet)
4. Distinction between Spark DataFrames and R data frames
Further pointers

http://spark.apache.org
http://www.r-project.org
http://www.ggplot2.org
https://cran.r-project.org/web/packages/magrittr
https://databricks.com/blog/2015/09/22/large-scale-topic-modeling-improvements-to-lda-on-spark.html
www.databricks.com

Office hour: 2:55 – 3:35 at Table A (O’Reilly Booth) / Databricks booth
Thank you