# Big Data Processing with Apache Spark and R

Hao Lin (Purdue University) Indianapolis, IN Oct. 20, 2015

Part of the slides are modified from Shivaram's slides

#### **About Me**

Graduate student @ Purdue ECE

Research interests includes Parallel/Distributed Computing, Data Analytics Infrastructure, Cloud Computing, Compilers; Major contributor of SparkR project

Data/Cloud Internship experience @ Huawei, Google and Facebook

#### **Outline**

#### **Overview**

SparkR R-RDD APIs

**DataFrame APIs** 

System Implementation & optimizations

Applications & Demos

**Future Directions** 

Q & A

#### **Motivation**

- Personal data for the internet
  - Query history
  - Click stream logging
  - > Tweets (8TB per day)
  - Facebook (500TB per day)
- Machine Generated Data
  - > Sensor networks
  - > Genome sequencing
  - > Physics experiments
  - Satellite imaging data
  - > System performance logs in data center



#### **Motivation**

Huge demand for data analysts, statisticians & data scientists

Traditional tools work on summary or sampled data

Properties of a good tool for large-scale data:

Usability: stick to traditional semantics



MB

- > Performance: latency, throughput, scalability, availability
- Fault Tolerance

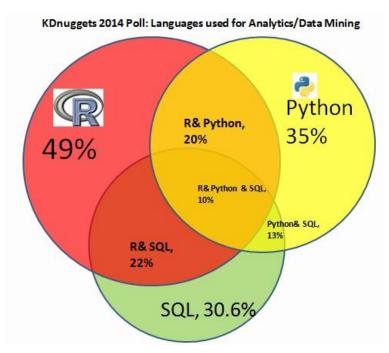
#### **Motivation**

R ranked top among Analytics/Data mining tools in all recent years Why R?:

- Data structures (list, array, DataFrame)
- Rich functionality in packages
- Graphic visualization
- Open source

#### Limitations of R?:

- x Single threaded
- x Limited memory



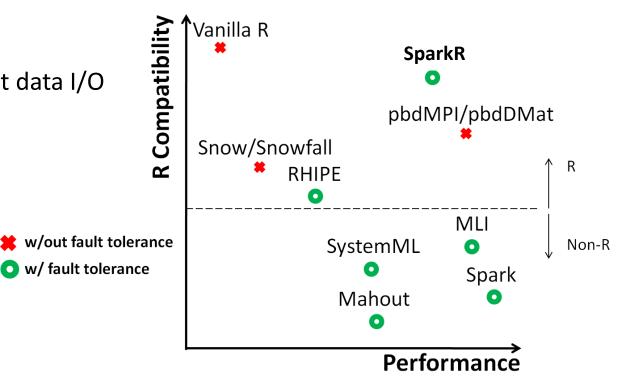
From: Survey by KDnuggets

http://www.kdnuggets.com/2014/08/four-main-languages-analytics-data-mining-data-science.html

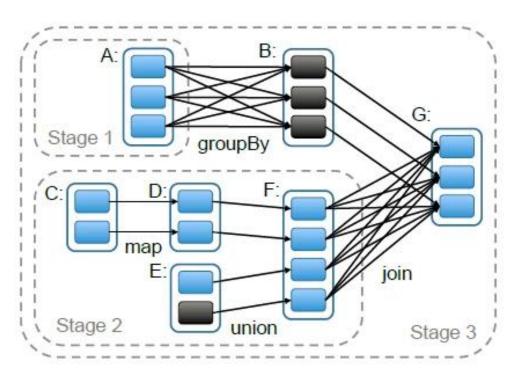
## Landscape

#### **Current Efforts:**

- R compatibility
- Performance & efficient data I/O
- Fault tolerance



## Spork Distributed Dataset (RDD) APIs



Developed at AMP lab, Berkeley Flexible Programming Model

- DAG-like execution planPerformance
- ✓ In-memory
- Good for iterative algorithmsResilient Data Sets
- Recover from loss and failures

From paper: Spark: Cluster Computing with Working Sets. HotCloud 2010.

## **SparkR**

Scalable Spork + Statistical

Packages

Flexible Plots

Part of the slides are modified from Shivaram's slides

#### **Software Stack**

Data Science Interface

SparkR

Data Processing

Engine

Spark

Cluster

Management

Mesos / YARN

**Data Storage** 

HDFS / HBase / Cassandra ...

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## **SparkR R-RDD APIs**

Considered as Distributed version of R List

```
Generation functions: textFile, parallelize, ...

Transformation functions: lapply, filter, sampleRDD, ...

Persistence function: cache, persist, ...

Action functions: reduce, collect, ...

Paired Value Shuffle functions: groupByKey, reduceByKey, ...

Binary Functions: unionRDD, cogroup, join, ...

Output Functions: saveAsTextFile, saveAsObjectFile, ...
```

## **R-RDD Example: Word Count**

```
library(SparkR)
lines <- textFile(sc, "hdfs://my_text_file")</pre>
```

## R-RDD Example: Word Count

```
library(SparkR)
lines <- textFile(sc, "hdfs://my_text_file")</pre>
words <- flatMap(lines,</pre>
                  function(line) {
                     strsplit(line, " ")[[1]]
wordCount <- lapply(words,</pre>
                      function(word) {
                        list(word, 1L)
```

## R-RDD Example: Word Count

```
library(SparkR)
lines <- textFile(sc, "hdfs://my text file")</pre>
words <- flatMap(lines,</pre>
                  function(line) {
                     strsplit(line, " ")[[1]]
wordCount <- lapply(words,</pre>
                      function(word) {
                        list(word, 1L)
counts <- reduceByKey(wordCount, "+", 2L)
output <- collect(counts)</pre>
```

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#### **Data Frames**

More structured data in Tables
Data source like CSV, JSON, JDBC, ...

Want to use your favorite package "dplyr"?

DataFrame type in SparkR

Embeded SQL in R

## Why R on Spark Data Frame?

Imposes a schema on RDDs

Query Optimizer, Code Gen

Rich DataSources API



















```
sqlCtx.table("people")\
.groupBy("name")\
.agg("name", avg("age")) \
.collect()
```

more sources at http://spark-packages.org/









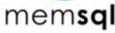












#### **DataFrame APIs**

Column access using '\$' or '[]' just like in R

dplyr-like DataFrame manipulation:

- filter
- groupBy
- summarize
- mutate

Access to external R packages that extend R syntax

#### **DataFrame APIs**

```
Filter -- Select some rows
filter(df, df$col1 > 0)
Project -- Select some columns
df$col1 or df["col"]
```

#### **DataFrame APIs**

```
Aggregate -- Group and Summarize data groupDF <- groupBy(df, df$col1)

agg(groupDF, sum(groupDF$col2), max(groupDF$col3))

Sort -- Sort data by a particular column

sortDF(df, asc(df$col1))
```

## **Column Average using RDD**

```
peopleRDD <- textFile(sc, "people.txt")</pre>
lines <- flatMap(peopleRDD,</pre>
                   function(line) {
                     strsplit(line, ", ")
                   })
ageInt <- lapply(lines,
                   function(line) {
                     as.numeric(line[2])
                   })
sum <- reduce(ageInt, function(x, y) {x+y})</pre>
avg <- sum / count(peopleRDD)</pre>
```

Name	Age
Andy	20
Hao	25

## **Column Average using DataFrame**

```
# JSON File contains two columns age, name
df <- jsonFile("people.json")
avg <- select(df, avg(df$age))</pre>
```

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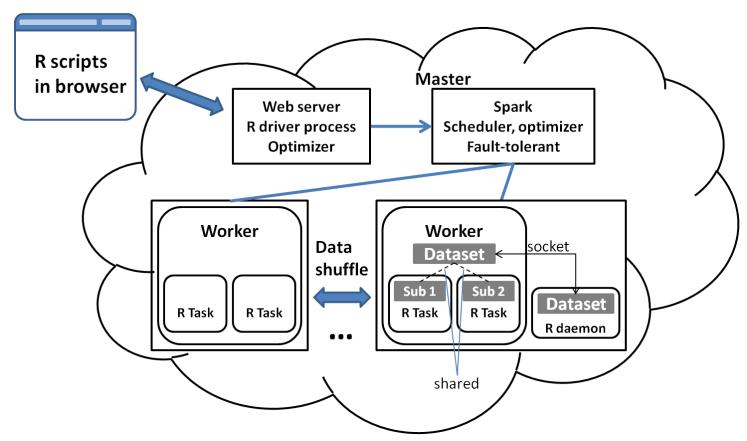
**DataFrame APIs** 

**System Implementation & optimizations** 

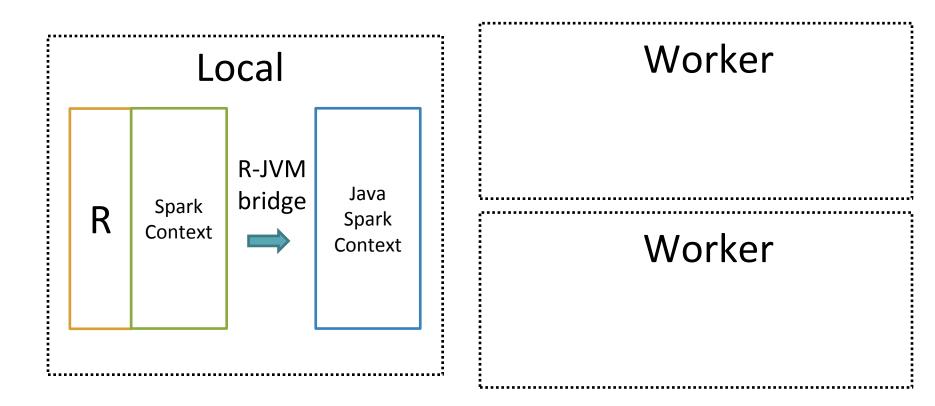
**Applications & Demos** 

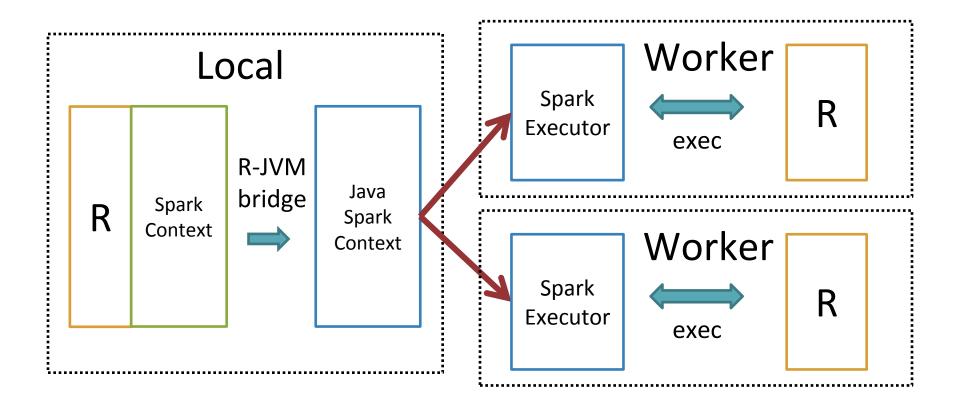
**Future Directions** 

Q & A



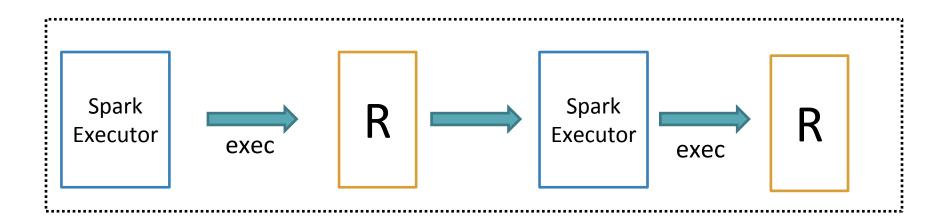
Worker Local Worker



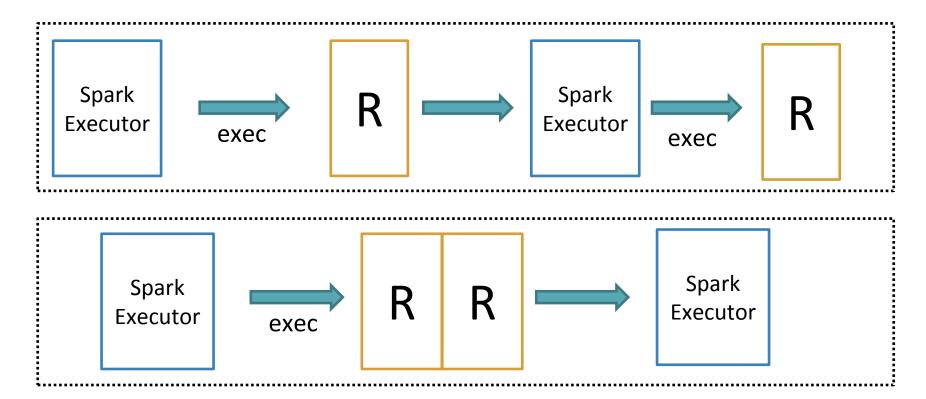


## **Pipelined RDD**

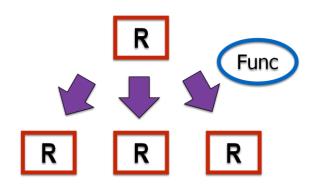
```
words <- flatMap(lines,...)
wordCount <- lapply(words,...)</pre>
```



## **Pipelined RDD**



## **Correctly Capture the Function Closure**



```
z <- 1
func1 <- function() {
    y <- 2
    func2 <- function(x) {
        x + y + z
    }</pre>
```

We need to ship the function closure to worker nodes

R has scoping rule to search for the values of free variables defined outside the function

We need to ship the functions together with the values of free variables in its environment

## lapply is Slow?

```
lapply(L, f) {
    len <- length(L)</pre>
    Lout <- alloc veclist(len)</pre>
    for(i in 1:len) {
         item <- L[[i]]
         Lout[[i]] <- f(item)
    return(Lout)
```

R is a interpreted language

Looping-over-data execution of lapply in R

Pick element one by one, and interpret f() many times

## Solution: Use lapplyPartition

```
grad.func <- function(yx) {</pre>
                                               Write vector
    y \leftarrow yx[1]
                                                function
    X < - yX
    x[1] \leftarrow 1 #modify the 1st column
                                                gradient <- function(partition) {</pre>
    logit \langle -1/(1 + \exp(-sum(theta*x))) \rangle
                                                     partition = partition[[1]]
    (v-logit) * x
                                                     Y <- partition[, 1]
                                                     X <- partition[, -1]</pre>
lapply(points, grad.func)
                                                     dot <- X %*% w
                                                     logit <- 1 / (1 + exp(-Y * dot))
                                                     grad <- t(X) %*% ((logit - 1) * Y)
                                                     list(grad)
```

lapplyPartition(points, gradient)

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### **Environment Setup**

Install R & RStudio: <a href="http://cran.r-project.org/">http://cran.r-project.org/</a>, <a href="http://cran.r-p

For latest Spark 1.4, SparkR is in the release <a href="https://github.com/apache/spark">https://github.com/apache/spark</a>
build SparkR by Maven, following <a href="https://github.com/apache/spark/tree/master/R">https://github.com/apache/spark/tree/master/R</a>
For Spark 1.3 or maybe before: download Spark from <a href="https://spark.apache.org/downloads.html">https://spark.apache.org/downloads.html</a>
Also install SparkR package in <a href="https://github.com/amplab-extras/SparkR-pkg">https://github.com/amplab-extras/SparkR-pkg</a>

Other packages like HDFS if necessary

## **Environment Setup**

Docker: <a href="https://registry.hub.docker.com/u/beniyama/sparkr-docker">https://registry.hub.docker.com/u/beniyama/sparkr-docker</a>

Amazon EC2: <a href="https://github.com/amplab-extras/SparkR-pkg/wiki/SparkR-on-EC2">https://github.com/amplab-extras/SparkR-pkg/wiki/SparkR-on-EC2</a>

Multiple node (Distributed mode) with HDFS

Install Hadoop: <a href="http://goo.gl/OXt1mC">http://goo.gl/OXt1mC</a>

Spark Standalone: <a href="https://spark.apache.org/docs/latest/spark-standalone.html">https://spark.apache.org/docs/latest/spark-standalone.html</a>

YARN: https://spark.apache.org/docs/latest/running-on-yarn.html

### **Quick Start**

#### Start RStudio

Desktop version: start RStudio application

Server version: open web browser with <your host>:8787



#### **Spark Context Initialization**

Setup SPARK\_HOME: alternatively we can store in ~/.Renviron s.t. it will not execute every time

```
# Set this to where Spark is installed
Sys.setenv(SPARK_HOME="/home/sparkr/workspace/spark")
# This line loads SparkR from the installed directory,
.libPaths(c(file.path(Sys.getenv("SPARK_HOME"), "R", "lib"), .libPaths()))
```

Load SparkR package: library(SparkR)

Init Spark Context: sc <- sparkR.init(master="local[n]")</pre>

### **Quick Start**

#### Also, you can always use terminal

```
For Spark 1.4: cd $SPARK_HOME; ./bin/sparkR --master "local[n]"
For Spark 1.3 with SparkR-pkg: cd SparkR-pkg; ./sparkR --master "local[n]"
Spark Context will automatically be created, call sc
```

# Pi Example

# **Logistic Regression**

# **Predicting Customer Behavior**

Demo from Chris Freeman from Alteryx

#### 3 datasets:

**Transactions** 

Demographic Info Per Customer

**DM Treatment Sample** 

How do we decide who to send the offer to?

# **Predicting Customer Behavior**

Demo from Chris Freeman from Alteryx

Use the DataFrame API to load, prepare and combine all 3 datasets and create training and estimation sets.

Use R's glm method to train a logistic regression model on the treatment sample

Profit!

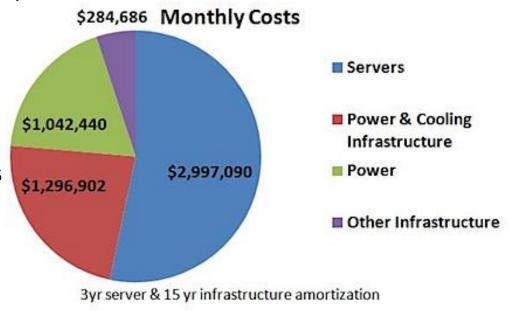
- Reduce number of physical resources used:
  - Capital expense (CAPEX)
  - > Operational expense (OPEX)

Increase resource utilization

Meet the SLO

Cost of 45,000 machines

From: James Hamilton http://perspectives.mvdirona.com/



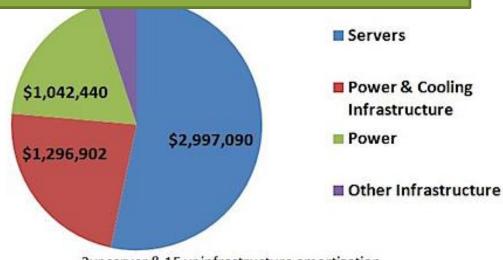
Historical system performance metrics can provide clues about efficient resource allocation.



 $\Box$   $\mathsf{R}$ 

Cost of 45,000 machines

From: James Hamilton http://perspectives.mvdirona.com/



3yr server & 15 yr infrastructure amortization

Historical system performance metrics can provide clues about efficient resource allocation.

Meet the SLO

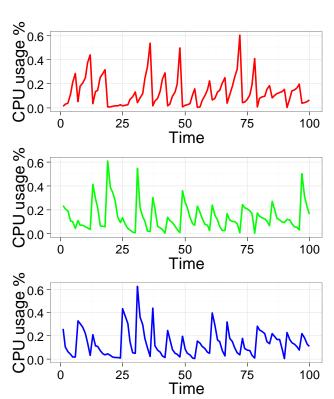
Servers

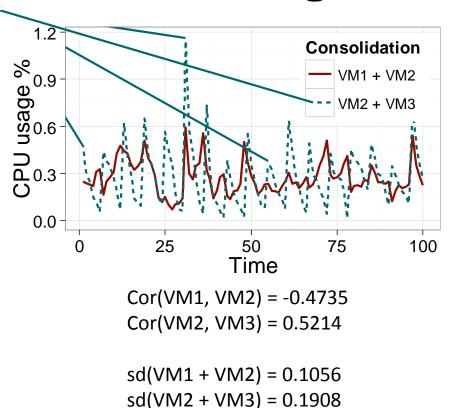
Machine generated workload metric history is a Large amount of data.

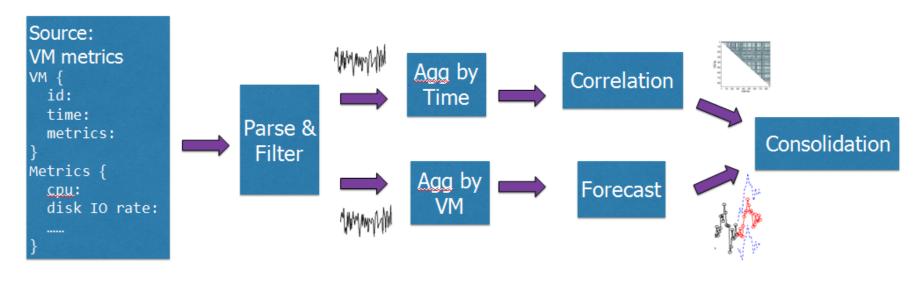
i ioiii. vaiiioo i iaiiiikoi

http://perspectives.mvdirona.com/

3yr server & 15 yr infrastructure amortization







Use SARIMA model in R

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### **Current Status and Community**

Originated in AMPLab

About 20 contributors AMPLab, Alteryx, Databricks, Intel

Now merged in Spark 1.4!



#### [SPARK-5654] Integrate SparkR #5096



### **Future Direction**

High-level APIs to do Machine learning Example: glm, kmeans

Pipelines with featurizers, learning

Tokenize → TF-IDF → LogisticRegression

Extended models, summary methods

#### **Future Direction**

APIs for Streaming, Time series analysis

Try a SparkR Streaming implementation:

https://github.com/hlin09/spark/tree/SparkR-streaming

https://github.com/hlin09/SparkR-pkg/tree/SparkR-streaming

Distributed matrix operations

<Your SparkR use case ?>

### Reference and Guide

Starter & RDD: <a href="https://github.com/amplab-extras/SparkR-">https://github.com/amplab-extras/SparkR-</a>

pkg/wiki/SparkR-Quick-Start

Data Frame: <a href="http://people.apache.org/~pwendell/spark-">http://people.apache.org/~pwendell/spark-</a>

releases/latest/sparkr.html

Chris's demo: <a href="https://github.com/cafreeman/Demo\_SparkR">https://github.com/cafreeman/Demo\_SparkR</a>

### **Other Resources**

#### Download VMware Player:

https://my.vmware.com/web/vmware/free#desktop\_end\_user\_computing/vmware\_player/7\_0

#### Download VM image:

http://web.ics.purdue.edu/~lin116/sparkr-ubuntu-1204-interface15.zip

#### Examples & Datasets:

http://web.ics.purdue.edu/~lin116/examples.tar.gz

http://web.ics.purdue.edu/~lin116/data.tar.gz

# Thanks! Questions?

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