# Difinity

Microsoft Data Platform
Business Intelligence Analytics
Conference

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www.difinity.co.nz





# Scalable Data Science with SparkR on HDInsight

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# Survey



# Agenda

- Spark Fundamentals
- Spark on HDInsight
- Overview of R and SparkR
- Machine Learning in SparkR
- UDFs in SparkR



# Spark Fundamentals









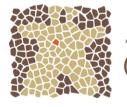












A P A C H E G I R A P H



### Difinity

# Apache Spark

a unified computing engine and a set of libraries for parallel data processing on computer clusters









Spark SQL

Spark Streaming Mllib (machine learning)

GraphX (graph)















Spark: The Definitive Guide, Matei Zaharia, Bill Chambers



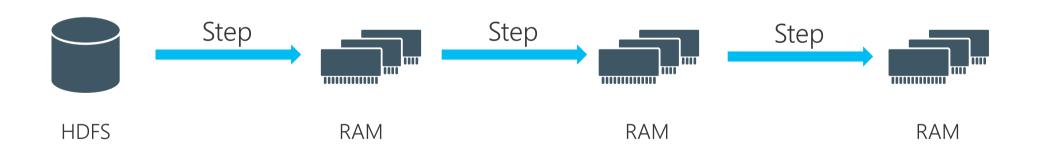


# Why Spark is fast



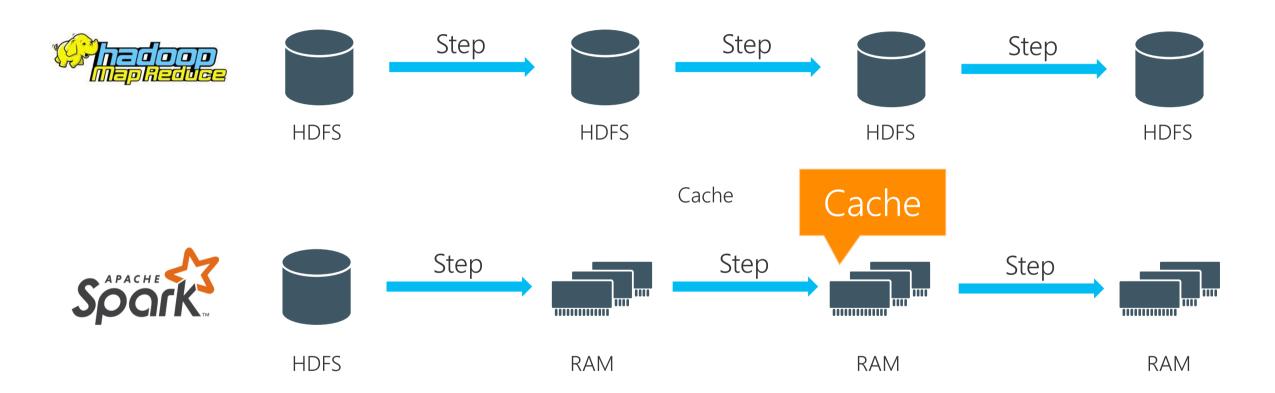






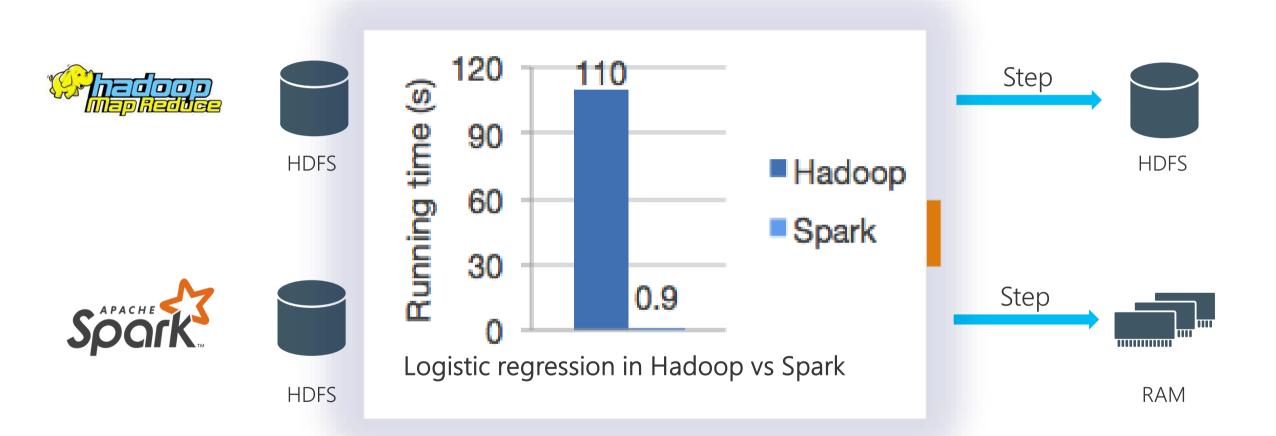


# Why Spark is fast





# Why Spark is fast



Source: http://spark.apache.org/





# Apache Spark: APIs

#### **RDDs**

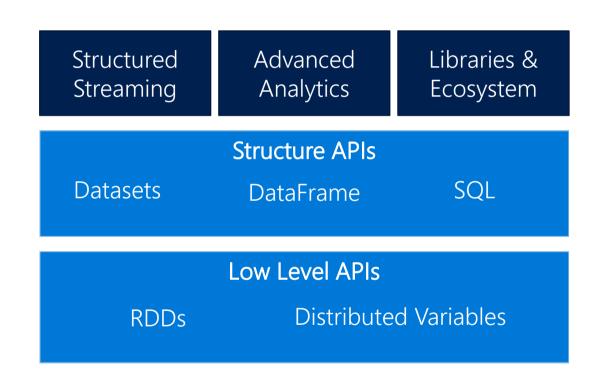
Core building block of data processing pipelines

#### **DataFrames**

High level APIs that take advantage of query optimizer

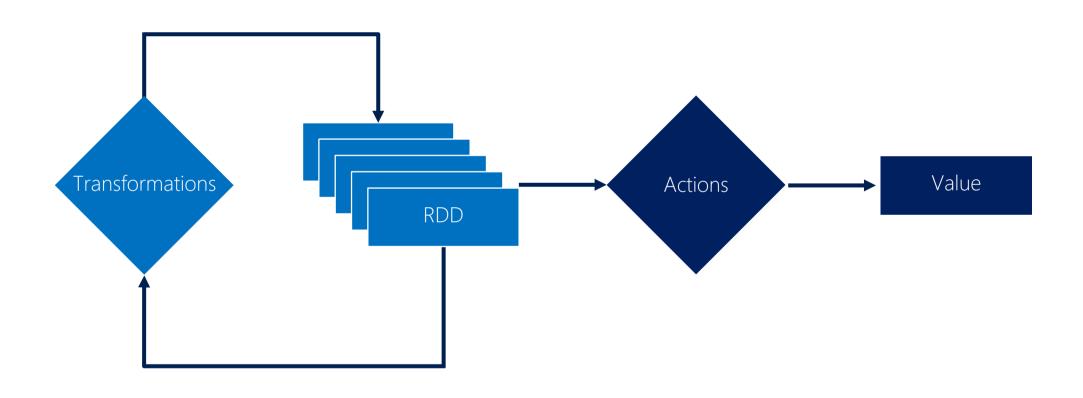
#### Datasets

Data Frames with user objects and custom code



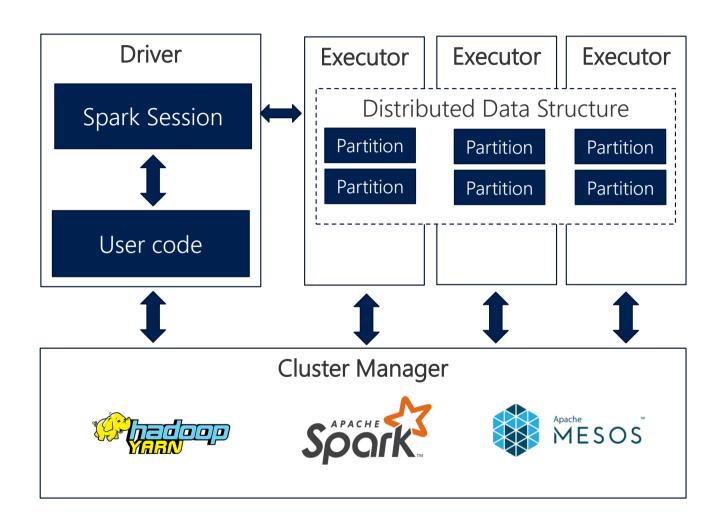


### Transformations and Actions





### Inside a Spark Application





# Spark on HDInsight

# Azure HDInsight

A Cloud Spark and Hadoop service for the Enterprise





- · Reliable with an industry leading SLA
- Enterprise-grade security and monitoring
- Productive platform for developers and scientists
- Cost effective cloud scale
- Integration with ISV applications
- Easy for administrators to manage
- 63% lower TCO than deploy your own Hadoop on-premises\*

\*IDC study "The Business Value and TCO Advantage of Apache Hadoop in the Cloud with Microsoft Azure HDInsight"





# Spark on Azure HDInsight

#### Fully Managed Service

Provision cluster with a click of a mouse

100% open source Apache Spark and Hadoop bits

Latest releases of Spark (1.6.3 and 2.1.1 are latest supported releases)

Fully supported by Microsoft and Hortonworks

99.9% Azure Cloud SLA

Certifications: PCI, ISO 27018, SOC, HIPAA, EU-MC

### Optimized for data exploration, experimentation & development

Jupyter/Zeppelin Notebooks (scala, python, automatic data visualizations)

IntelliJ/Eclipse plugins (job submission, remote debugging)

ODBC connector for Power BI, Tableau, Qlik, SAP, Excel, etc.





# R Server on HDInsight

#### Spark cluster

- ... with a Microsoft R Server edge node
- ... R installed across the nodes
- ... enterprise-scale R analytics
- ... multi-threaded math libraries



# Demo

Hello, HDInsight

# SparkR

### R



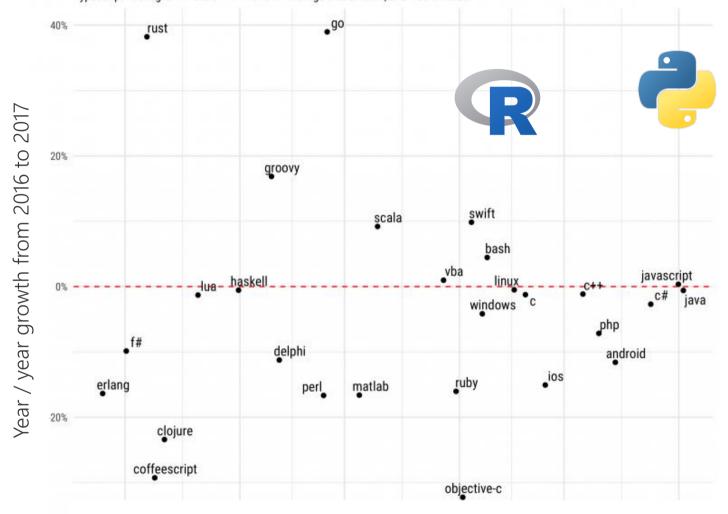
- Statistical programming language
- Rich ecosystem of packages in CRAN
  - · 10,000+ packages
- Powerful data visualization libraries
- Created by Ross Ihaka and Robert Gentleman at the University of Auckland, New Zealand
  - · Initial version in 1995. Based on S language (1975)
- · Interpreted and single threaded





#### Year over year growth in traffic to programming languages/platforms

Comparing question views in January-September of 2016 and 2017, in World Bank high-income countries. TypeScript had a growth rate of 134% and an average size of .38%; and was omitted.



Average % of Stack Overflow visits across the two years (log scale)

Source: The Impressive Growth of R by David Robinson, https://stackoverflow.blog/2017/10/10/impressive-growth-r/

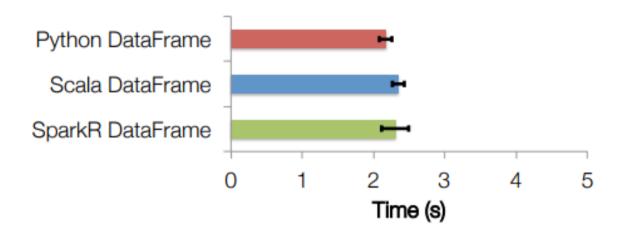




### SparkR

- Light-weight R frontend to Spark
- Exposes Spark's DataFrame API
- MLlib bindings
- R package
- REPL (sparkR)



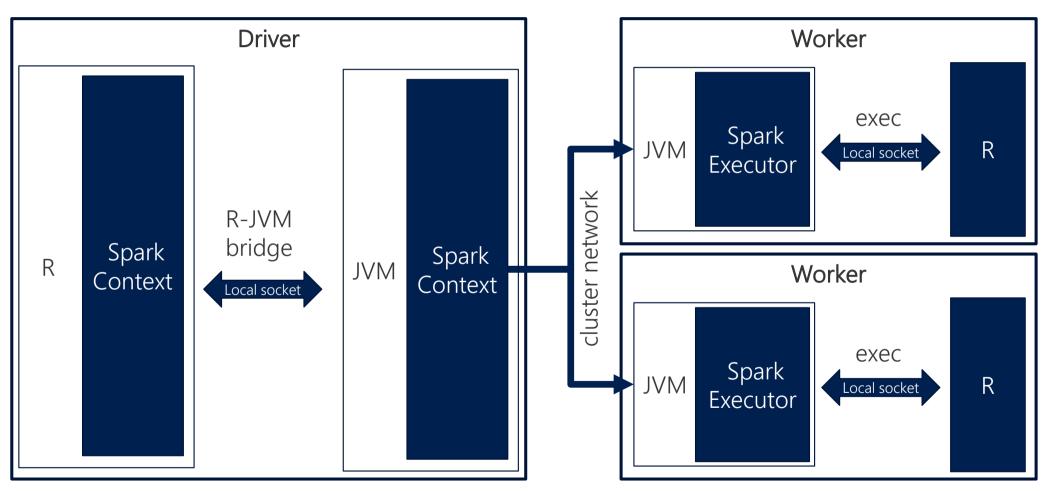


Source: http://people.csail.mit.edu/matei/papers/2016/sigmod\_sparkr.pdf





# SparkR Architecture



Source: http://people.csail.mit.edu/matei/papers/2016/sigmod\_sparkr.pdf





# Creating a SparkSession

```
# Load SparkR library
library(SparkR, lib.loc = "/path/to/package")
# Start a spark session
sparkR.session()
# Inspect spark session
sparkR.conf()
# Stop a spark session
sparkR.session.stop()
```



# Creating a SparkSession



### Reading and writing data

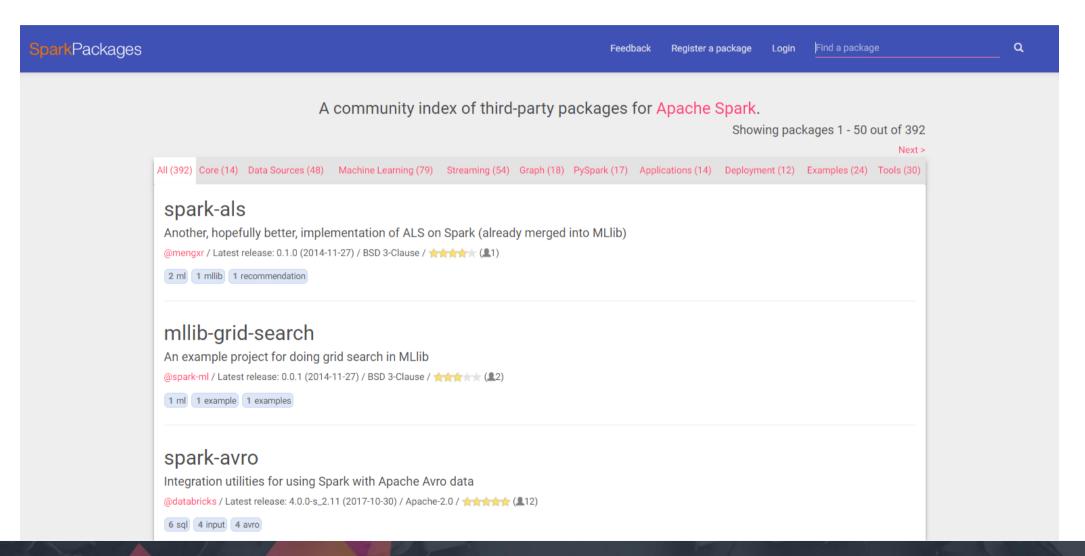
```
read.json()
read.orc()
read.parquet()
read.text()
read.jdbc()
read.df()
```

# corresponding write.format() methods





# More at spark-packages.org







# Converting between R and Spark

```
# R DataFrame → Spark DataFrame
sdf <- createDataFrame(rdf)</pre>
```

```
# Spark DataFrame → R DataFrame rdf <- collect(sdf)
```





# SparkSQL in SparkR

```
# Query a hive table
sdf <- sql("SELECT * FROM hiveTable")</pre>
```

```
# On an existing Spark DataFrame
createOrReplaceTempView(sdf, "myTable")
sdf2 <- sql("SELECT * FROM myTable")</pre>
```





# Manipulating Data in SparkR

```
# Transformations
filter()
select()
join()
groupBy()
pivot()
explode()
summarize()
sample()
distinct()
arrange()
# and more ...
```

```
# Actions
collect()
head()
first()
nrow()
take()
```

# Seattle Public Library

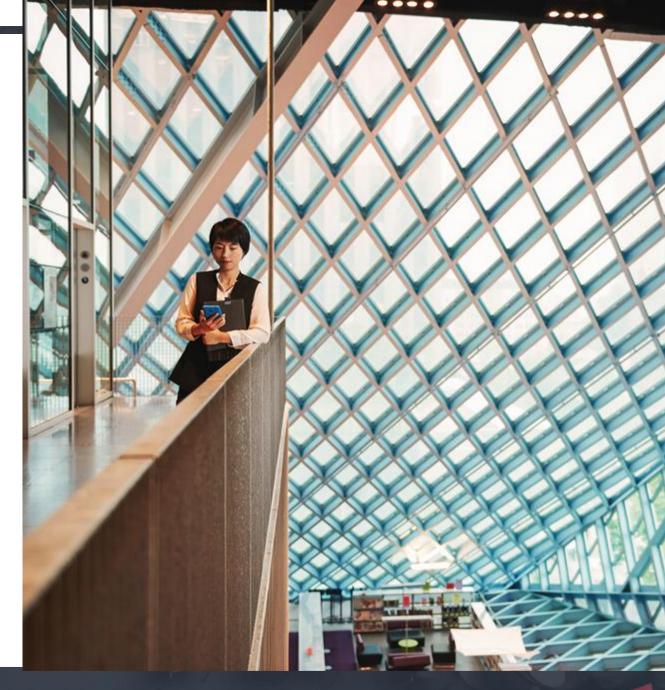
Predicting daily borrowed book types and collections

#### Given:

- Checkout history
- Library collection

Data source:

https://data.seattle.gov



# Demo

Hello, SparkR



### Window Functions

- Operates on a group of rows while returning a single value per row
- Aggregate functions are n to 1, Window functions are n to n
- Great for rolling operations





### Window Functions

# Demo

SparkR: Window Functions and Feature Engineering

# Machine Learning w/ SparkR



## Machine Learning models in SparkR

#### Classification

- Logistic Regression
- Multilayer Perceptron (MLP)
- Naive Bayes
- Linear Support Vector Machine (Spark 2.2 only)

#### Regression

- Accelerated Failure Time (AFT) Survival Model
- Generalized Linear Model (GLM)
- · Isotonic Regression

#### Tree

- Gradient Boosted Trees for Regression and Classification
- · Random Forest for Regression and Classification

#### Clustering

- Bisecting k-means (Spark 2.2 only)
- Gaussian Mixture Model (GMM)
- K-Means
- Latent Dirichlet Allocation (LDA)

#### Collaborative Filtering

- Alternating Least Squares (ALS)
- Frequent Pattern Mining (Spark 2.2 only)
- FP-growth (Spark 2.2 only)

#### **Statistics**

Kolmogorov-Smirnov Test





## Split data

```
# Split into Train and Test
sdfSplit <- sdfCheckoutsWeekly %>%
    randomSplit(weights = c(7, 3), seed = 123)
sdfTrain <- sdfSplit[[1]]
sdfTest <- sdfSplit[[2]]</pre>
```



### Example: Random Forest

```
# Fit model
model <- spark.randomForest(train,
    label ~ feature1 + feature2, type =
    "regression", maxDepth = 5, maxBins = 16,
    numTrees = 20, seed = 10)</pre>
```

```
# Predictions
predictions <- predict(model, newdata = test)</pre>
```





### Example: Random Forest

```
# Fit model
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```

#### # Predictions

predictions <- predict(model, newdata = test)</pre>





### Model Persistence in SparkR

```
# Save model
write.ml(model, path = "/model/path")
# Load model
model <- read.ml(path = "/model/path")</pre>
```



# Demo

SparkR: Machine Learning



## Parallelizing Native R

User Defined Functions (UDFs)

By partition – dapply or dapplyCollect

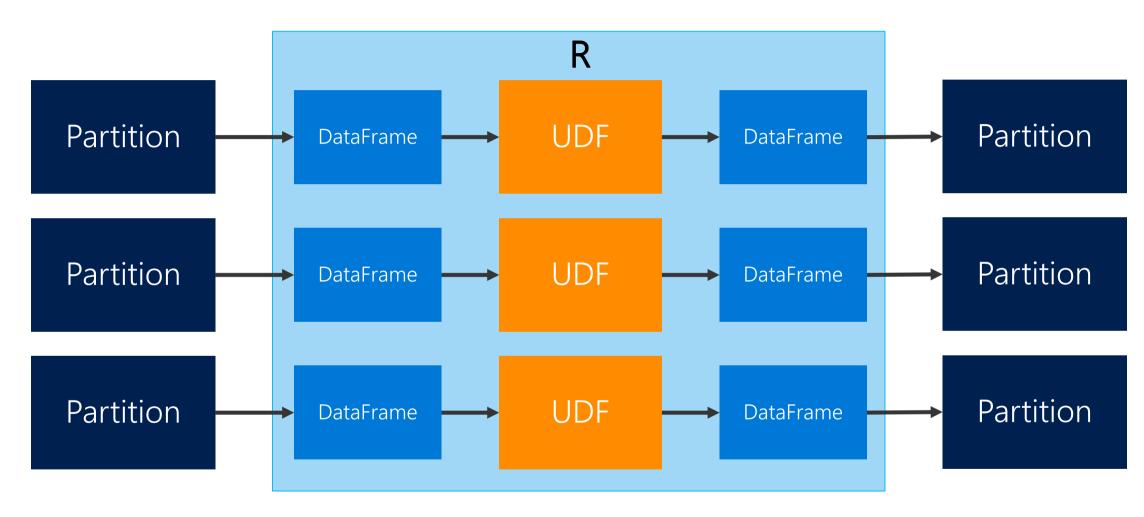
By group — gapply or gapplyCollect

Distributed R – lapply





# Parallel Processing By Partition (dapply)



Source: Scalable Data Science with SparkR: Spark Summit East talk by Felix Cheung





## Example: dapply

```
# Select specific columns
sdfSubCars <- select(sdfCars, "model", "mpg")</pre>
# Define schema
schema <- "model STRING, mpg DOUBLE, kmpg DOUBLE"
# Use dapply
out <- dapply(sdfSubCars, function(x) {</pre>
     x \leftarrow cbind(x, x pg * 1.61)
}, schema)
```

# Demo

Installing R packages using script actions



# Challenges with SparkR

- Can be hard to debug ML jobs
- · Not as mature as the Python, Java, Scala interfaces
- Can't call directly feature transformers/extractors
  - Currently, fixed pipelines
  - Not all Spark MLlib functions are directly exposed/ported
- Collects can be slow





### More to checkout...

### R on Spark

- ScaleR / RxSpark (Machine Learning Server / R Server)
- Sparklyr
- H2O R

#### Managed Spark on Azure

Azure Databricks



Q & A



# Thank you!

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