# SPARKLYR:

# RECAP, UPDATES AND USE CASES JAVIER LURASCHI

**SPARK SUMMIT 2017** 

#### **SCHEDULE**

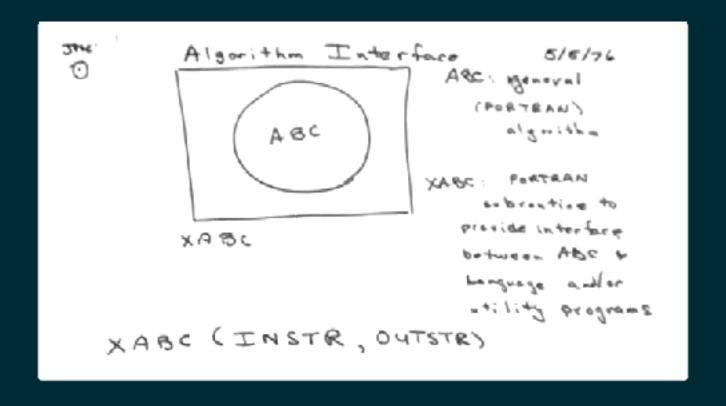
2:00-2:30 sparklyr: Recap and Updates

2:40-3:10 sparklyr: Architecture and Use Cases



#### S - LANGUAGE FOR STATS COMPUTING

Designed at Bell Laboratories by John Chambers, where computing was done by calling Fortran subroutines.



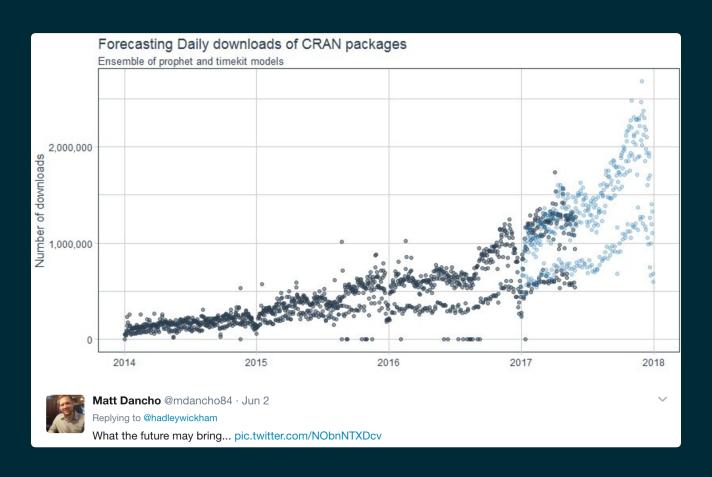
## S - LANGUAGE FOR STATS COMPUTING

"S is great but serious data analysis will always be done in Fortran" - Bell Labs Management

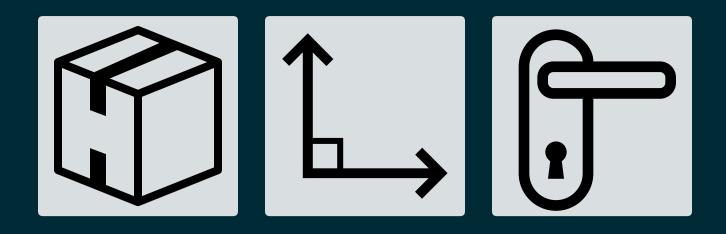


#### R - MODERN S

R community is noted for its active package contributions. CRAN R's package manager with ~10K packages.



#### R - DESIGN PRINCIPLES



- 1. Everything that exists is an object
- 2. Everything that happens is a function call
- 3. R is built on interfaces to many (R and non-R) algorithms

#### **SPARKLYR - R INTERFACE FOR SPARK**

```
library(sparklyr)
                                        # Load sparklyr
spark install()
                                        # Install Apache Spark
sc <- spark connect(master = "local") # Connect to local instance</pre>
library(dplyr)
                                        # Data Manipulation Grammar
mtcars tbl <- copy to(sc, mtcars)</pre>
                                        # Copy mtcars into Spark
                                        # Count records
count(mtcars tbl)
                                 # Perform linear regression
ml linear regression(mtcars tbl,
 response = "mpg",
                                        # Response vector
 features = c("wt", "cyl"))
                                        # Features for the model fit
library(DBI)
                                       # R Database Interface
dbGetQuery(sc, "SELECT * FROM mtcars") # Run SQL query in Spark
invoke(spark context(sc), "version") # Run sc.version in Scala
compile package jars()
                                        # Compile Scala code
```

# **UPDATES**

#### SPARKLYR 0.4

KICKOFF: April, 2016

**ANNOUNCED:** June, 2016

**RELEASED: September, 2016** 

**CLOUDERA CERTIFIED: October, 2016** 

**NEW FEATURES**: Install, connection, backend, data, DataFrame, DBI, dplyr, MLlib, extensions

#### SPARKLYR 0.5

**RELEASED: January 2017** 

MINOR: 0.5.2, 0.5.3, 0.5.4 and 0.5.5

**NEW CONNECTIONS:** Gateway, Livy and Databricks

IMPROVEMENTS: MLIib, DataFrame, compatibility and dplyr

## SPARKLYR 0.6 (DEVEL)

**RELEASED: Soon** 

**NEW FEATURES: Distributed R** 

IMPROVEMENTS: Data, dplyr, databases, DataFrames, MLIib, broom, compatibility, connections, extensions and backend

#### SPARK-INSTALL

#### Cross-platform installer for Apache Spark.

```
library(sparkinstall)
spark_install(version = "1.6.2") # Install Spark from R

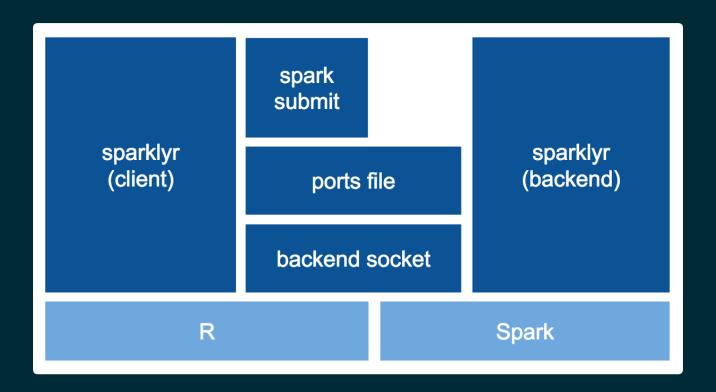
from spark_install import *
spark_install(version = "1.6.2") # Install Spark from Python
```

"This project provides a cross-platform installer for Apache Spark designed to use system resources efficiently under a common API. This initial version commes with support for R and Python that arose from a collaboration between RStudio and Microsoft" - github.com/rstudio/spark-install

# **ARCHITECTURE**

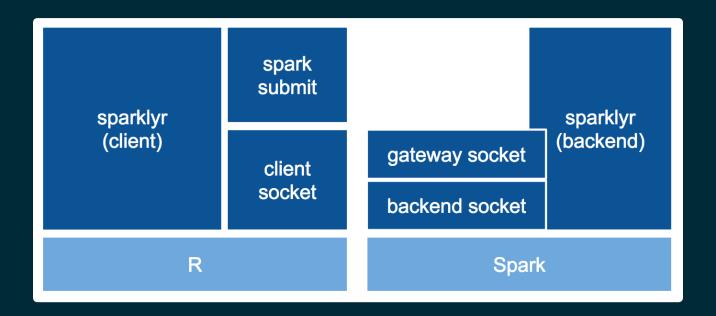
## **BACKEND**

#### SPARKLYR 0.4



#### **GATEWAY**

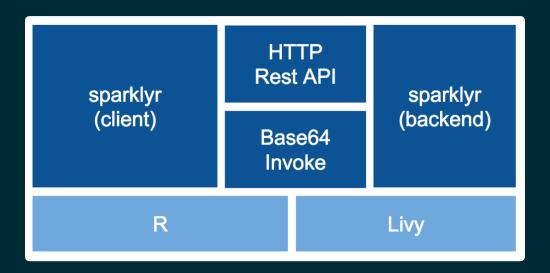
#### SPARKLYR 0.5



Replace ports file with gateway socket



#### SPARKLYR 0.5

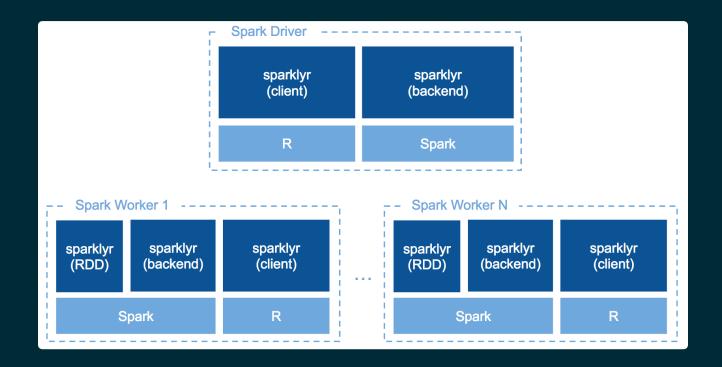


```
var sparklyrRetVar_0 = LivyUtils.invokeFromBase64(
   "AAAAHm9yZy5hcGFjaGUuc3BhcmsuU3BhcmtDb250" +
   "ZXh0AAAAAAEAAAAMZ2V0T3JDcmVhdGUAAAAAAA=="
)
```

#### Implement Livy connections

### WORKER

#### **SPARKLYR 0.6**



Implement R Workers

# **USE CASES**

#### ANALYSIS WITH SQL AND DPLYR

```
delay <- flights_tbl %>%
   group_by(tailnum) %>%
   summarise(count = n(), dist = mean(distance), delay = mean(arr_delay)
   filter(count > 20, dist < 2000, !is.na(delay)) %>%
   collect

# plot delays
library(ggplot2)
ggplot(delay, aes(dist, delay)) +
   geom_point(aes(size = count), alpha = 1/2) +
   geom_smooth() +
   scale_size_area(max_size = 2)
```

```
library(DBI)
dbGetQuery(sc, "SELECT * FROM flights LIMIT 100")
```

#### MACHINE LEARNING WITH MLLIB

```
# transform our data set, and then partition into 'training', 'test'
partitions <- mtcars_tbl %>%
   filter(hp >= 100) %>%
   mutate(cyl8 = cyl == 8) %>%
   sdf_partition(training = 0.5, test = 0.5, seed = 1099)

# fit a linear model to the training dataset
partitions$training %>%
   ml_linear_regression(response = "mpg", features = c("wt", "cyl"))
```

#### MACHINE LEARNING WITH RSPARKLING

#### EXTENSION BY H20 (Navdeep Gill)

```
library(rsparkling)
library(sparklyr)
library(dplyr)
library(h2o)
sc <- spark connect(master = "local")</pre>
mtcars h2o <- as h2o frame(sc, mtcars tbl,</pre>
                             strict version check = FALSE)
h2o.glm(x = c("wt", "cyl"),
        y = "mpq",
        training frame = mtcars h2o,
        lambda search = TRUE)
h2o flow(sc, strict version check = FALSE)
```

#### GRAPHFRAMES WITH SPARKLYGRAPHS

#### **EXTENSION BY Kevin Kuo**

```
spark_disconnect(sc)
```

```
library(sparklygraphs)
library(sparklyr)
library(dplyr)
sc <- spark connect(master = "local", version = "2.1.0")</pre>
highschool tbl <- copy to(sc, ggraph::highschool, "highschool")
# create a table with unique vertices using dplyr
vertices tbl <- sdf bind rows(</pre>
  highschool tbl %>% distinct(from) %>% transmute(id = from),
  highschool tbl %>% distinct(to) %>% transmute(id = to)
# create a table with <source, destination> edges
edges tbl <- highschool tbl %>% transmute(src = from, dst = to)
# calculate PageRank over the highschool dataset
qf qraphframe(vertices tbl, edges tbl) %>%
  qf pagerank(reset prob = 0.15, max iter = 10L, source id = "1")
```

## **DISTRIBUTED EXECUTION**

#### **SPARKLYR 0.6**

```
spark_apply(highschool_tbl, function(x) {
   x + rgamma(1, 2)
})
```

# THANK YOU!

# **QUESTIONS?**

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spark\_disconnect(sc)