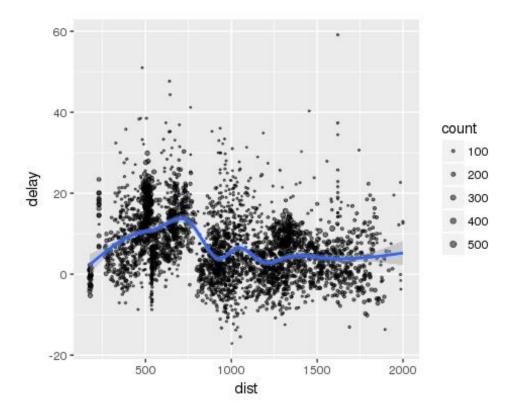
sparklyr

sparklyr — R interface for Apache Spark 의 설치와 사용법

CentOS 6.7

```
사전 준비
sudo yum -y install libcurl-devel
install.packages("sparklyr" , repos = 'http://cran.nexr.com')
library(sparklyr)
# spark_install(version = "1.6.2")
if (nchar(Sys.getenv("SPARK_HOME")) < 1) {</pre>
  Sys.setenv(SPARK_HOME = "/home/goodmit/spark")
}
sc <- spark_connect(master = "local")</pre>
Reading Data
install.packages("dplyr", repos = "http://cran.nexr.com" )
install.packages("nycflights13", repos = "http://cran.nexr.com" )
install.packages("Lahman" , repos = "http://cran.nexr.com" )
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
##
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(nycflights13)
iris_tbl <- copy_to(sc, iris)</pre>
## The following columns have been renamed:
## - 'Sepal.Length' => 'Sepal_Length' (#1)
## - 'Sepal.Width' => 'Sepal Width'
                                       (#2)
## - 'Petal.Length' => 'Petal_Length' (#3)
## - 'Petal.Width' => 'Petal_Width'
```

```
flights tbl <- copy to(sc, nycflights13::flights, "flights")
batting_tbl <- copy_to(sc, Lahman::Batting, "batting")</pre>
src tbls(sc)
## [1] "batting" "flights" "iris"
Using dplyr
# filter by departure delay
flights_tbl %>% filter(dep_delay == 2)
             query [?? x 19]
## Source:
## Database: spark connection master=local[1] app=sparklyr local=TRUE
##
##
       year month
                    day dep_time sched_dep_time dep_delay arr_time
##
      <int> <int> <int>
                            <int>
                                           <int>
                                                      <dbl>
                                                               <int>
## 1
       2013
                1
                      1
                              517
                                             515
                                                          2
                                                                 830
                                                          2
## 2
       2013
                1
                      1
                              542
                                             540
                                                                 923
## 3
                                                          2
       2013
                1
                      1
                              702
                                             700
                                                                1058
       2013
                1
                              715
                                                          2
## 4
                      1
                                             713
                                                                 911
                                                          2
## 5
       2013
                1
                      1
                              752
                                             750
                                                                1025
## 6
       2013
                1
                      1
                              917
                                             915
                                                          2
                                                                1206
                                                          2
## 7
       2013
                1
                      1
                              932
                                             930
                                                                1219
## 8
       2013
                1
                      1
                             1028
                                             1026
                                                          2
                                                                1350
## 9
       2013
                1
                      1
                             1042
                                            1040
                                                          2
                                                                1325
## 10 2013
                1
                      1
                             1231
                                            1229
                                                          2
                                                                1523
## # ... with more rows, and 12 more variables: sched_arr_time <int>,
       arr delay <dbl>, carrier <chr>, flight <int>, tailnum <chr>,
       origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>, hour <dbl>,
## #
## #
       minute <dbl>, time_hour <dbl>
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()
install.packages('ggplot2', repos = 'http://cran.nexr.com')
# 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)
## `geom_smooth()` using method = 'gam'
```



spark + R 의 장점

• Spark 에서 병렬처리가 가능하도록 구현한 머신러닝 LIB 을 사용하고, 결과를 R 의 시각화 패키지로 보여줌.

K-Means Clustering

```
kmeans_model <- iris_tbl %>%
  select(Petal_Width, Petal_Length) %>%
  ml_kmeans(centers = 3)
# print our model fit
print(kmeans_model)
## K-means clustering with 3 clusters
##
## Cluster centers:
     Petal_Width Petal_Length
        0.246000
                     1.462000
## 1
## 2
        2.037500
                     5.595833
## 3
        1.342308
                     4.269231
##
## Within Set Sum of Squared Errors = 31.37136
```

```
predicted <- sdf_predict(kmeans_model, iris_tbl) %>% collect
table(predicted$Species, predicted$prediction)
##
##
                0 1 2
##
    setosa
                50 0 0
##
    versicolor 0 2 48
##
    virginica
                0 46 4
# plot cluster membership
sdf_predict(kmeans_model) %>%
 collect() %>%
 ggplot(aes(Petal_Length, Petal_Width)) +
 geom_point(aes(Petal_Width, Petal_Length, col = factor(prediction + 1)),
             size = 2, alpha = 0.5) +
 geom_point(data = kmeans_model$centers, aes(Petal_Width, Petal_Length),
             col = scales::muted(c("red", "green", "blue")),
             pch = 'x', size = 12) +
 scale_color_discrete(name = "Predicted Cluster",
                       labels = paste("Cluster", 1:3)) +
 labs(
   x = "Petal Length",
   y = "Petal Width",
   title = "K-Means Clustering",
    subtitle = "Use Spark.ML to predict cluster membership with the iris data
set."
)
```

K-Means Clustering

Use Spark.ML to predict cluster membership with the iris dataset.

