

# R Notebook

## Parametros:

```
Measure = F1 measure
Columns = learner
Performance = holdout_measure_residual
Filter keys = sampling, weight_space, ruspool
Filter values = FALSE, FALSE, FALSE
```

```
library("scmamp")
library(dplyr)
```

## Tratamento dos dados

Carregando data set compilado

```
ds = read.csv("/home/rodrigo/Dropbox/UNICAMP/IC/estudo_cost_learning/SummaryResults/summary_compilation.
summary(ds)
```

```
##           learner      weight_space
## classif.ksvm      :17100  Mode :logical
## classif.randomForest:17100 FALSE:41040
## classif.xgboost    :17100  TRUE :10260
##                                     NA's :0
##
##
##
##           measure      sampling      ruspool
## Accuracy              :10260  ADASYN:10260  Mode :logical
## Area under the curve   :10260  FALSE :30780  FALSE:41040
## F1 measure              :10260  SMOTE :10260  TRUE :10260
## G-mean                 :10260                      NA's :0
## Matthews correlation coefficient:10260
##
##
## tuning_measure  holdout_measure  holdout_measure_residual
## Min.      :-0.1277  Min.      :-0.2120  Min.      :-0.4658
## 1st Qu.: 0.5924  1st Qu.: 0.3114  1st Qu.: 0.1648
## Median : 0.9624  Median : 0.8193  Median : 0.5192
## Mean      : 0.7570  Mean      : 0.6469  Mean      : 0.5099
## 3rd Qu.: 0.9965  3rd Qu.: 0.9879  3rd Qu.: 0.8636
## Max.      : 1.0000  Max.      : 1.0000  Max.      : 1.0000
## NA's      :1761    NA's      :1761    NA's      :1761
## iteration_count      dataset      imba.rate
## Min.      :1      abalone      : 900  Min.      :0.0010
## 1st Qu.:1      adult      : 900  1st Qu.:0.0100
## Median :2      bank      : 900  Median :0.0300
## Mean      :2      car      : 900  Mean      :0.0286
## 3rd Qu.:3      cardiotocography-10clases: 900  3rd Qu.:0.0500
## Max.      :3      cardiotocography-3clases : 900  Max.      :0.0500
```

```
## NA's :1761 (Other) :45900
```

Filtrando pela metrica

```
ds = filter(ds, measure == params$measure)
```

Filtrando o data set

```
if(params$filter_keys != 'NULL' && !is.null(params$filter_keys)){  
  ds = filter_at(ds, .vars = params$filter_keys, .vars_predicate = any_vars(. == params$filter_values))  
}
```

```
summary(ds)
```

```
##           learner      weight_space  
## classif.ksvm      :3420  Mode :logical  
## classif.randomForest:3420 FALSE:8208  
## classif.xgboost    :3420  TRUE :2052  
##                   NA's :0  
##  
##  
##  
##           measure      sampling      ruspool  
## Accuracy           : 0 ADASYN:2052  Mode :logical  
## Area under the curve : 0 FALSE :6156  FALSE:8208  
## F1 measure           :10260 SMOTE :2052  TRUE :2052  
## G-mean              : 0           NA's :0  
## Matthews correlation coefficient: 0  
##  
##  
## tuning_measure  holdout_measure  holdout_measure_residual  
## Min. :0.0000  Min. :0.0000  Min. :0.0000  
## 1st Qu.:0.1667 1st Qu.:0.0000 1st Qu.:0.0187  
## Median :0.7363 Median :0.3704 Median :0.2328  
## Mean :0.5997 Mean :0.4199 Mean :0.3418  
## 3rd Qu.:0.9922 3rd Qu.:0.8000 3rd Qu.:0.6582  
## Max. :1.0000 Max. :1.0000 Max. :1.0000  
## NA's :354 NA's :354 NA's :354  
## iteration_count      dataset      imba.rate  
## Min. :1      abalone : 180  Min. :0.0010  
## 1st Qu.:1      adult : 180  1st Qu.:0.0100  
## Median :2      bank : 180  Median :0.0300  
## Mean :2      car : 180  Mean :0.0286  
## 3rd Qu.:3      cardiotocography-10clases: 180 3rd Qu.:0.0500  
## Max. :3      cardiotocography-3clases : 180 Max. :0.0500  
## NA's :354 (Other) :9180
```

Computando as médias das iteracoes

```
ds = group_by(ds, learner , weight_space , measure , sampling , ruspool , dataset , imba.rate)  
ds = summarise(ds, tuning_measure = mean(tuning_measure), holdout_measure = mean(holdout_measure),  
               holdout_measure_residual = mean(holdout_measure_residual))  
  
ds = as.data.frame(ds)
```

Criando dataframe

```

# Dividindo o ds em n, um para cada técnica
splited_df = ds %>% group_by_at(.vars = params$columns) %>% do(vals = as.data.frame(.)) %>% select(vals)

# Juntando cada uma das partes horizontalmente em um data set
df_tec_wide = do.call("cbind", splited_df)

# Renomeando duplicacao de nomes
colnames(df_tec_wide) = make.unique(colnames(df_tec_wide))

# Selecionando apenas as medidas da performance escolhida
df_tec_wide_residual = select(df_tec_wide, matches(paste("^", params$performance, "$|", params$performance)))

# Renomeando colunas
new_names = NULL
for(i in (1:length(splited_df))){
  id = toString(sapply(splited_df[[i]][1, params$columns], as.character))
  new_names = c(new_names, id)
}
colnames(df_tec_wide_residual) = new_names

# Verificando a dimensao do df
dim(df_tec_wide_residual)

## [1] 1140    3

# Removendo linhas com NA's
df_tec_wide_residual = na.omit(df_tec_wide_residual)

# Renomeando a variavel
df = df_tec_wide_residual

summary(df)

```

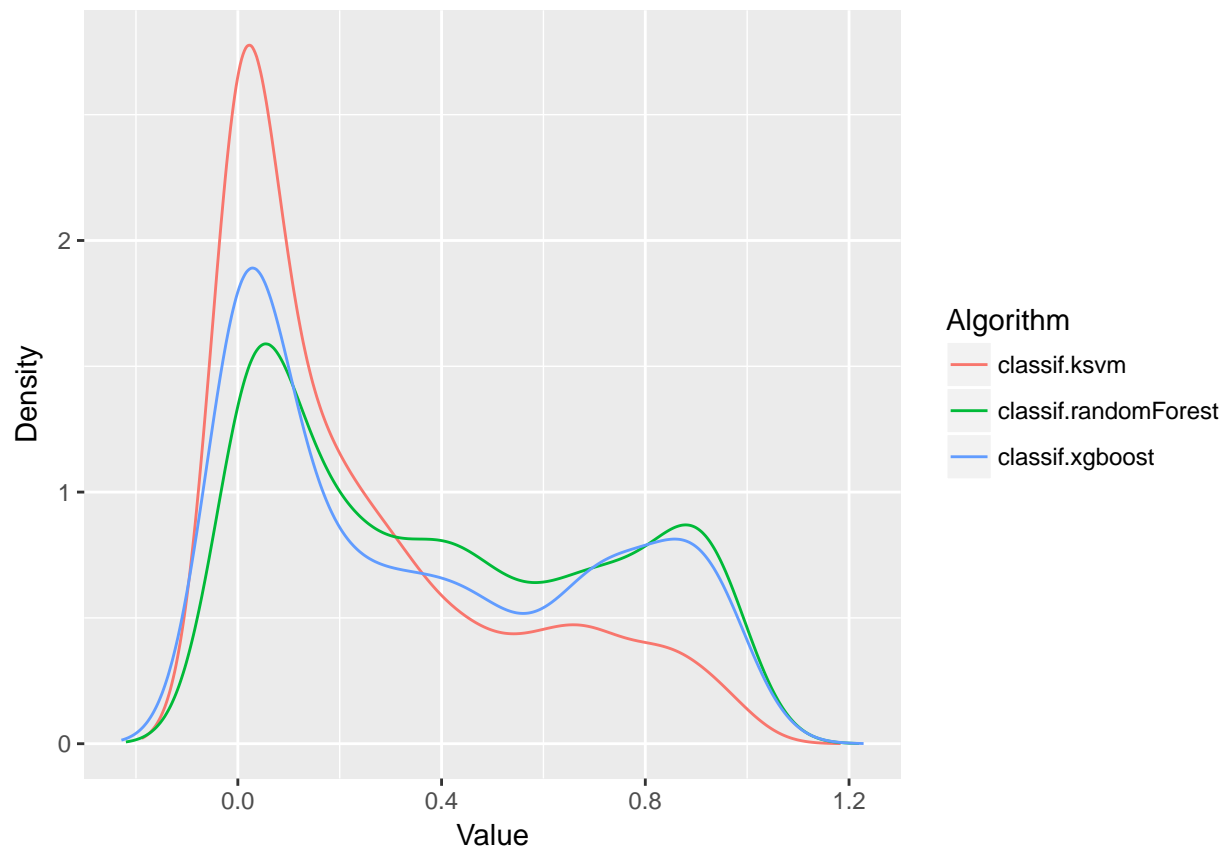
```

##   classif.ksvm      classif.randomForest classif.xgboost
## Min.   :0.000000   Min.   :0.00000   Min.   :0.00000
## 1st Qu.:0.006536   1st Qu.:0.08345   1st Qu.:0.02283
## Median :0.132462   Median :0.35936   Median :0.27846
## Mean   :0.243768   Mean   :0.40447   Mean   :0.36430
## 3rd Qu.:0.404813   3rd Qu.:0.70750   3rd Qu.:0.69450
## Max.   :0.994911   Max.   :1.00000   Max.   :1.00000

```

## Fazendo teste de normalidade

```
plotDensities(data = df)
```



## Testando as diferencas

```
friedmanTest(df)

##
## Friedman's rank sum test
##
## data: df
## Friedman's chi-squared = 293.84, df = 2, p-value < 2.2e-16
```

## Testando as diferencas par a par

```
test <- nemenyiTest (df, alpha=0.05)
abs(test$diff.matrix) > test$statistic

##      classif.ksvm classif.randomForest classif.xgboost
## [1,]      FALSE              TRUE      TRUE
## [2,]       TRUE              FALSE      TRUE
## [3,]       TRUE              TRUE      FALSE
```

## Plotando grafico de Critical Difference

```
result = tryCatch({  
  plotCD(df, alpha=0.05, cex = 0.35)  
}, error = function(e) {})
```

