

# R Notebook

## Parametros:

Measure = Matthews correlation coefficient  
Columns = sampling, weight\_space, underbagging  
Performance = tuning\_measure  
Filter keys = imba.rate  
Filter values = 0.03

```
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.  
ds = filter(ds, learner != "classif.rusboost")  
summary(ds)
```

```
##           learner      weight_space  
## classif.ksvm      :17100  Mode :logical  
## classif.randomForest:17100 FALSE:41040  
## classif.rusboost   :    0  TRUE :10260  
## classif.xgboost    :17100  NA's :0  
##  
##  
##  
##           measure      sampling      underbagging  
## 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.: 0.6911  1st Qu.: 0.4001  1st Qu.: 0.1994  
## Median : 0.9700  Median : 0.8571  Median : 0.5581  
## Mean   : 0.7903  Mean   : 0.6718  Mean   : 0.5298  
## 3rd Qu.: 0.9975  3rd Qu.: 0.9900  3rd Qu.: 0.8755  
## Max.   : 1.0000  Max.   : 1.0000  Max.   : 1.0000  
## NA's   :1077    NA's   :1077    NA's   :1077  
## 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      cardiocography-10clases: 900 3rd Qu.:0.0500
## Max. :3      cardiocography-3clases : 900 Max. :0.0500
## NA's :1077 (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)){
  dots = paste0(params$filter_keys, " == '",params$filter_values, "'")
  ds = filter_(ds, .dots = dots)
}
```

```
summary(ds)
```

```
##           learner      weight_space
## classif.ksvm      :990  Mode :logical
## classif.randomForest:990 FALSE:2376
## classif.rusboost   : 0  TRUE :594
## classif.xgboost    :990  NA's :0
##
##
##
##           measure      sampling      underbagging
## Accuracy           : 0  ADASYN: 594  Mode :logical
## Area under the curve : 0  FALSE :1782 FALSE:2376
## F1 measure          : 0  SMOTE : 594  TRUE :594
## G-mean              : 0              NA's :0
## Matthews correlation coefficient:2970
##
##
## tuning_measure      holdout_measure      holdout_measure_residual
## Min. : -0.05673  Min. : -0.1757  Min. : -0.4658
## 1st Qu.: 0.33347  1st Qu.: 0.0000  1st Qu.: 0.0391
## Median : 0.83196  Median : 0.5030  Median : 0.2116
## Mean : 0.66187  Mean : 0.4753  Mean : 0.3111
## 3rd Qu.: 0.98596  3rd Qu.: 0.8126  3rd Qu.: 0.5286
## Max. : 1.00000  Max. : 1.0000  Max. : 1.0000
## NA's :48      NA's :48      NA's :48
## iteration_count      dataset      imba.rate
## Min. :1      abalone : 45  Min. :0.03
## 1st Qu.:1      adult : 45  1st Qu.:0.03
## Median :2      annealing : 45  Median :0.03
## Mean :2      arrhythmia : 45  Mean :0.03
## 3rd Qu.:3      balance-scale: 45  3rd Qu.:0.03
## Max. :3      bank : 45  Max. :0.03
## NA's :48      (Other) :2700
```

Computando as médias das iteracoes

```
ds = group_by(ds, learner , weight_space , measure , sampling , underbagging , 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] 198 5
```

```
# Renomeando a variavel
df = df_tec_wide_residual

head(df)
```

```
## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, FALSE, TRUE
## 1 0.9202676 0.02820236 0.11727958
## 2 0.9501846 0.11635362 0.20937678
## 3 0.9664565 0.17638584 0.29672610
## 4 0.8624084 0.00000000 0.11395088
## 5 1.0000000 0.95818163 0.77180093
## 6 0.9705260 0.05285858 0.09049135
## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## 1 0.01746777 0.9235585
## 2 0.14047268 0.9528485
## 3 0.17638584 0.9648315
## 4 0.00000000 0.7663273
## 5 0.95818163 1.0000000
## 6 0.05285858 0.9688113
```

```
summary(df)
```

```
## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, FALSE, TRUE
## Min. :0.6105 Min. : -0.004855 Min. :0.006784
## 1st Qu.:0.9610 1st Qu.: 0.086965 1st Qu.:0.222084
## Median :0.9896 Median : 0.516278 Median :0.421194
## Mean :0.9690 Mean : 0.474868 Mean :0.432180
## 3rd Qu.:0.9965 3rd Qu.: 0.812064 3rd Qu.:0.645409
## Max. :1.0000 Max. : 1.000000 Max. :1.000000
```

```
## NA's :6          NA's :2          NA's :2
## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## Min. : -0.002485 Min. :0.6164
## 1st Qu.: 0.096142 1st Qu.:0.9635
## Median : 0.493988 Median :0.9898
## Mean : 0.472091 Mean :0.9706
## 3rd Qu.: 0.825359 3rd Qu.:0.9974
## Max. : 1.000000 Max. :1.0000
## NA's :2          NA's :4
```

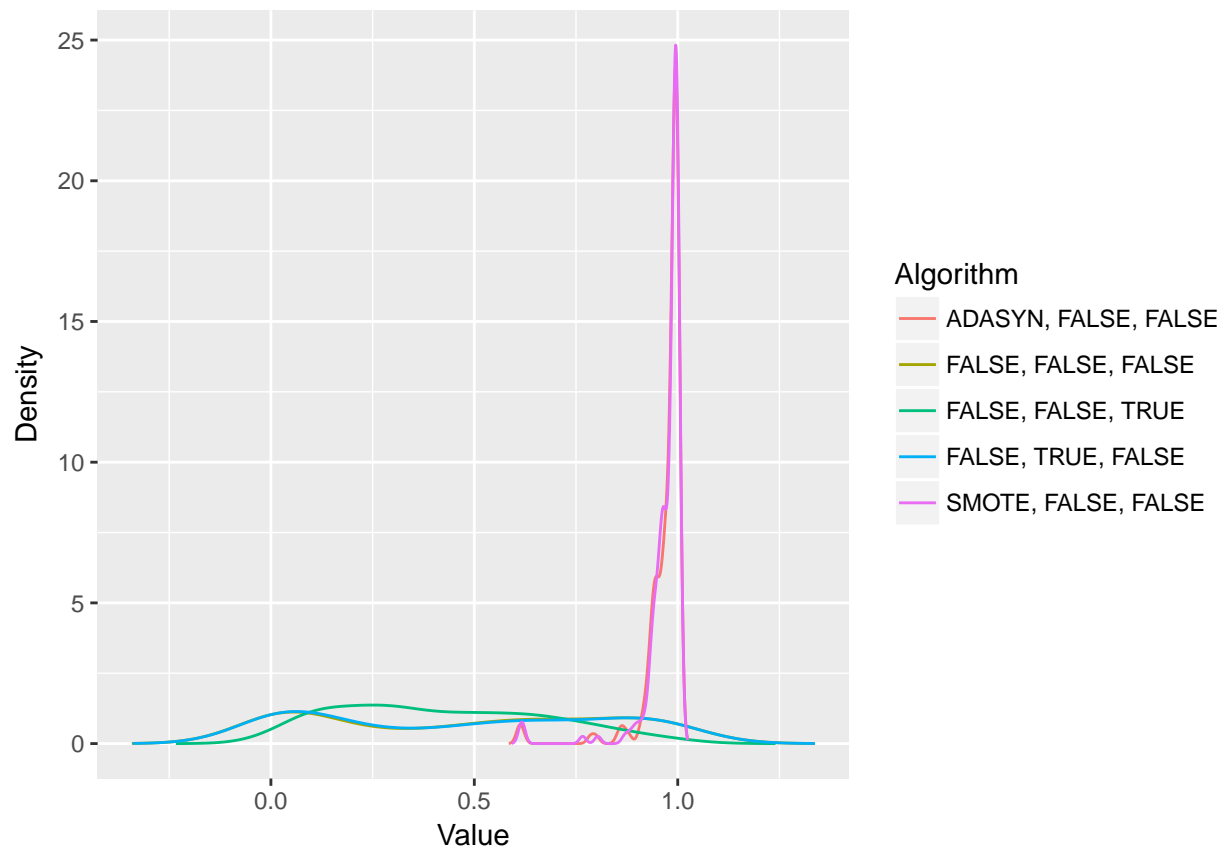
## Verificando a média de cada coluna selecionada

```
for(i in (1:dim(df)[2])){
  print(paste("Media da coluna ", colnames(df)[i], " = ", mean(df[,i], na.rm = TRUE), sep=""))
}

## [1] "Media da coluna ADASYN, FALSE, FALSE = 0.968984422721258"
## [1] "Media da coluna FALSE, FALSE, FALSE = 0.474868280000896"
## [1] "Media da coluna FALSE, FALSE, TRUE = 0.432180390603151"
## [1] "Media da coluna FALSE, TRUE, FALSE = 0.472090883367927"
## [1] "Media da coluna SMOTE, FALSE, FALSE = 0.97062535664133"
```

## Fazendo teste de normalidade

```
plotDensities(data = na.omit(df))
```



## Testando as diferencas

```
friedmanTest(df)

##
##  Friedman's rank sum test
##
## data:  df
## Friedman's chi-squared = 486.01, df = 4, p-value < 2.2e-16
```

## Testando as diferencas par a par

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

##      ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, FALSE, TRUE
## [1,]          FALSE          TRUE          TRUE
## [2,]          TRUE          FALSE          FALSE
## [3,]          TRUE          FALSE          FALSE
## [4,]          TRUE          FALSE          FALSE
## [5,]          FALSE          TRUE          TRUE
##      FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## [1,]          TRUE          FALSE
```

```
## [2,]          FALSE          TRUE
## [3,]          FALSE          TRUE
## [4,]          FALSE          TRUE
## [5,]           TRUE          FALSE
```

## Plotando os ranks

```
print(colMeans(rankMatrix(df)))
```

```
## ADASYN, FALSE, FALSE  FALSE, FALSE, FALSE  FALSE, FALSE, TRUE
##           1.638889           3.823232           4.103535
##  FALSE, TRUE, FALSE  SMOTE, FALSE, FALSE
##           3.772727           1.661616
```

## Plotando grafico de Critical Difference

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

