R. Notebook

Parametros:

Mean :2

car

```
Measure = Matthews correlation coefficient

Columns = sampling, weight_space, underbagging

Performance = holdout_measure

Filter keys = imba.rate

Filter values = 0.01

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)
##
                                weight_space
                   learner
                       :17100
                                Mode :logical
##
   classif.ksvm
   classif.randomForest:17100
                                FALSE:41040
   classif.rusboost
                                TRUE: 10260
                      :
##
   classif.xgboost
                       :17100
                                NA's :0
##
##
##
##
                               measure
                                             sampling
                                                          underbagging
##
   Accuracy
                                   :10260
                                           ADASYN:10260
                                                          Mode :logical
                                           FALSE :30780
##
  Area under the curve
                                   :10260
                                                          FALSE: 41040
## F1 measure
                                           SMOTE :10260
                                                          TRUE :10260
                                   :10260
##
   G-mean
                                   :10260
                                                          NA's :0
   Matthews correlation coefficient:10260
##
##
##
  tuning_measure
##
                     holdout_measure
                                      holdout_measure_residual
  Min.
         :-0.1277
                     Min. :-0.2120
                                            :-0.4658
##
                                      Min.
  1st Qu.: 0.6911
                     1st Qu.: 0.4001
                                      1st Qu.: 0.1994
## Median : 0.9700
                     Median : 0.8571
                                      Median : 0.5581
                     Mean : 0.6718
## Mean : 0.7903
                                      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
                                              900
                   bank
                                                    Median :0.0300
```

900

Mean :0.0286

```
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
                        :600
                               Mode :logical
## classif.randomForest:600
                               FALSE: 1440
                        : 0
  classif.rusboost
                               TRUE: 360
   classif.xgboost
                        :600
                               NA's :0
##
##
##
##
                                              sampling
                                                          underbagging
                                measure
                                            ADASYN: 360
##
   Accuracy
                                    :
                                       0
                                                          Mode :logical
   Area under the curve
                                       0
                                           FALSE :1080
                                                          FALSE: 1440
  F1 measure
                                       0
                                            SMOTE: 360
                                                          TRUE :360
##
                                                          NA's :0
   G-mean
  Matthews correlation coefficient: 1800
##
##
##
                                        holdout_measure_residual
##
  tuning_measure
                       holdout_measure
## Min. :-0.00646
                            :-0.1370
                                              :-0.06817
                      Min.
                                        Min.
  1st Qu.: 0.23261
                      1st Qu.: 0.0000
                                       1st Qu.: 0.02011
   Median : 0.82014
                      Median : 0.3764
                                       Median: 0.19200
          : 0.64070
                             : 0.4285
                                                : 0.29498
## Mean
                      Mean
                                       Mean
  3rd Qu.: 0.99730
                       3rd Qu.: 0.8152
                                         3rd Qu.: 0.49996
## Max.
          : 1.00000
                      Max.
                              : 1.0000
                                        Max.
                                                : 1.00000
## NA's
           :69
                       NA's
                              :69
                                         NA's
                                                :69
## iteration_count
                                         dataset
                                                       imba.rate
                                                          :0.01
## Min. :1
                   abalone
                                             : 45
                                                     Min.
## 1st Qu.:1
                    adult.
                                               45
                                                     1st Qu.:0.01
## Median :2
                    bank
                                                     Median:0.01
                                                45
                                                          :0.01
## Mean
         :2
                    car
                                                45
                                                    Mean
## 3rd Qu.:3
                    cardiotocography-10clases:
                                                45
                                                     3rd Qu.:0.01
## Max.
                                                            :0.01
          :3
                    cardiotocography-3clases:
                                                45
                                                     Max.
## NA's
          :69
                    (Other)
                                             :1530
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)
```

900

900

:45900

3rd Qu.:0.0500

:0.0500

Max.

3rd Qu.:3

:3

:1077

(Other)

Max.

NA's

cardiotocography-10clases:

cardiotocography-3clases :

```
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, "$|", params$performa
# 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] 120
# Renomeando a variavel
df = df_tec_wide_residual
head(df)
    ADASYN, FALSE, FALSE FALSE, FALSE, FALSE, FALSE, TRUE
##
## 1
            -0.014499516
                                -0.002681154
                                                     0.03648651
## 2
            -0.009081137
                                 0.011520780
                                                     0.11251890
## 3
            -0.004917732
                                 0.000000000
                                                     0.09881969
## 4
             0.463579851
                                 0.858510613
                                                     1.00000000
## 5
             0.000000000
                                                     0.00000000
                                 0.636257514
## 6
             0.191940287
                                 0.735548937
                                                     0.46374461
##
   FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## 1
          -0.004028968
                              -0.015024224
## 2
           0.009851304
                               0.012608807
## 3
           0.000000000
                              -0.004539511
## 4
                               0.605069238
           0.858510613
## 5
           0.636257514
                               0.166038918
## 6
           0.735548937
                               0.191940287
summary(df)
## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, TRUE
         :-0.0145
                      Min.
                              :-0.005742 Min.
                                                  :-0.02588
## 1st Qu.: 0.0000
                        1st Qu.: 0.000000
                                            1st Qu.: 0.14800
## Median : 0.3365
                        Median : 0.465389
                                            Median : 0.27777
                        Mean : 0.465610
## Mean : 0.4247
                                            Mean : 0.35146
## 3rd Qu.: 0.8023
                        3rd Qu.: 0.872325
                                            3rd Qu.: 0.50993
## Max. : 1.0000
                        Max. : 1.000000 Max. : 1.00000
```

```
## NA's :10
                     NA's :3
                                        NA's :1
## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## Min. :-0.005031
                     Min.
                           :-0.01672
## 1st Qu.: 0.000000 1st Qu.: 0.00000
## Median : 0.467017
                     Median: 0.38977
## Mean
         : 0.460559 Mean : 0.44164
## 3rd Qu.: 0.858511
                     3rd Qu.: 0.83054
## Max. : 1.000000
                           : 1.00000
                     Max.
## NA's
         :4
                     NA's
                            :5
```

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 = 0.424664743950473"

## [1] "Media da coluna FALSE, FALSE, FALSE = 0.465610453584806"

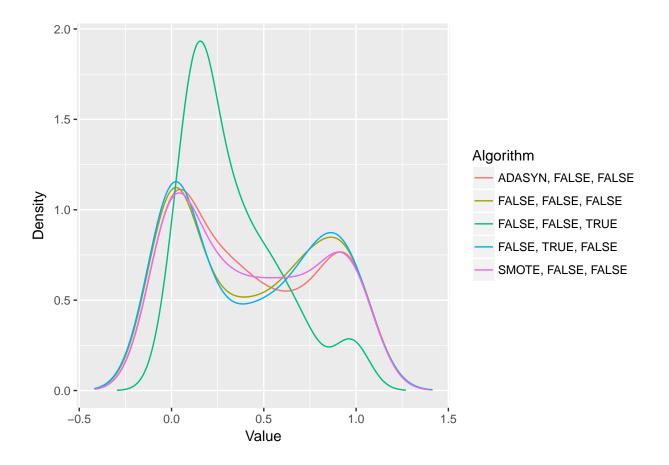
## [1] "Media da coluna FALSE, FALSE, TRUE = 0.351458120128611"

## [1] "Media da coluna FALSE, TRUE, FALSE = 0.460559366861479"

## [1] "Media da coluna SMOTE, FALSE, FALSE = 0.441642007742059"
```

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 = 2.2783, df = 4, p-value = 0.6847
```

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

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

Plotando os ranks

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

## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE, FALSE, TRUE
## 3.045833 2.891667 3.125000

## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## 2.879167 3.058333
```

Plotando grafico de Critical Diference

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

