

R Notebook

Parametros:

Measure = Matthews correlation coefficient
Columns = sampling, weight_space, underbagging
Performance = tuning_measure
Filter keys = imba.rate
Filter values = 0.001

```
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      :600 Mode :logical
## classif.randomForest:600 FALSE:1440
## classif.rusboost   : 0 TRUE :360
## classif.xgboost    :600 NA's :0
##
##
##
##          measure      sampling      underbagging
## Accuracy          : 0 ADASYN: 360 Mode :logical
## Area under the curve : 0 FALSE :1080 FALSE:1440
## F1 measure          : 0 SMOTE : 360 TRUE :360
## G-mean              : 0 NA's :0
## Matthews correlation coefficient:1800
##
##
## tuning_measure      holdout_measure      holdout_measure_residual
## Min. : -0.00646 Min. : -0.1370 Min. : -0.06817
## 1st Qu.: 0.23511 1st Qu.: 0.0000 1st Qu.: 0.02210
## Median : 0.81997 Median : 0.3764 Median : 0.19355
## Mean : 0.64034 Mean : 0.4305 Mean : 0.29627
## 3rd Qu.: 0.99727 3rd Qu.: 0.8152 3rd Qu.: 0.49996
## Max. : 1.00000 Max. : 1.0000 Max. : 1.00000
## NA's :54 NA's :54 NA's :54
## iteration_count      dataset      imba.rate
## Min. :1 abalone : 45 Min. :0.001
## 1st Qu.:1 adult : 45 1st Qu.:0.001
## Median :2 bank : 45 Median :0.001
## Mean :2 car : 45 Mean :0.001
## 3rd Qu.:3 cardiocography-10clases: 45 3rd Qu.:0.001
## Max. :3 cardiocography-3clases : 45 Max. :0.001
## NA's :54 (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)
```

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] 120 5
```

```
# Renomeando a variavel
df = df_tec_wide_residual
```

```
head(df)
```

```
## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, FALSE, TRUE
## 1 0.9711549 -0.00331073 0.03775174
## 2 0.9840938 0.03130559 0.09124314
## 3 0.9976886 0.00000000 0.07291374
## 4 1.0000000 0.84986516 0.83204223
## 5 1.0000000 0.53692615 0.42292522
## 6 0.9973459 0.66437344 0.86395957
## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## 1 -0.002739151 0.9694672
## 2 0.043297878 0.9861198
## 3 0.000000000 0.9973734
## 4 0.849865157 1.0000000
## 5 0.536926146 1.0000000
## 6 0.664373439 0.9973574
```

```
summary(df)
```

```
## ADASYN, FALSE, FALSE FALSE, FALSE, FALSE FALSE, FALSE, TRUE
## Min. :0.6780 Min. : -0.003311 Min. :0.02596
## 1st Qu.:0.9907 1st Qu.: 0.055346 1st Qu.:0.15045
## Median :0.9969 Median : 0.463649 Median :0.27765
## Mean :0.9899 Mean : 0.452866 Mean :0.35267
## 3rd Qu.:0.9992 3rd Qu.: 0.777142 3rd Qu.:0.47872
## Max. :1.0000 Max. : 1.000000 Max. :1.00000
```

```
## NA's      :8          NA's      :1
## FALSE, TRUE, FALSE SMOTE, FALSE, FALSE
## Min.      :-0.002739 Min.      :0.8777
## 1st Qu.: 0.055339 1st Qu.:0.9927
## Median : 0.465670 Median :0.9974
## Mean      : 0.452086 Mean      :0.9933
## 3rd Qu.: 0.778992 3rd Qu.:0.9995
## Max.      : 1.000000 Max.      :1.0000
## NA's      :2          NA's      :7
```

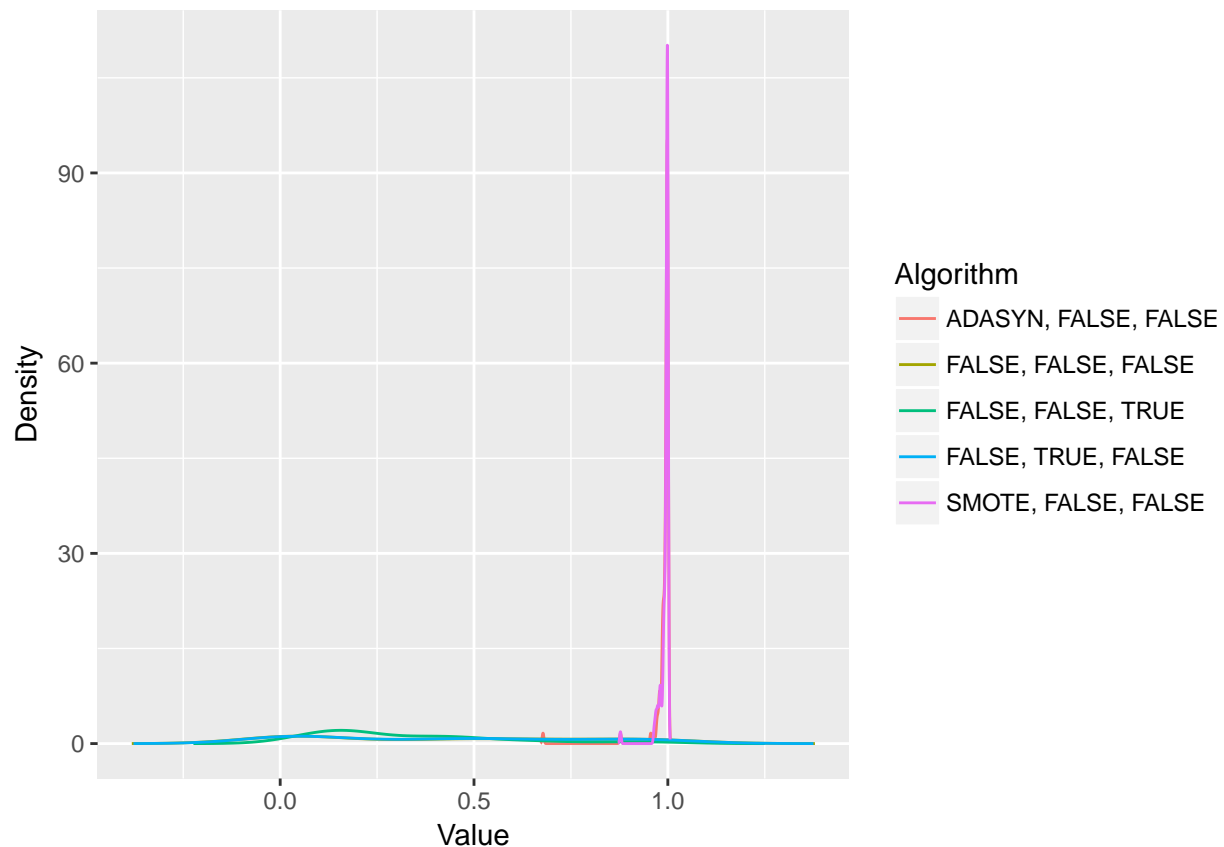
Verificando a média de cada coluna selecionada

```
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 = 0.989940626546064"
## [1] "Media da coluna FALSE, FALSE, FALSE = 0.452866152386498"
## [1] "Media da coluna FALSE, FALSE, TRUE = 0.352665664413952"
## [1] "Media da coluna FALSE, TRUE, FALSE = 0.452085947913061"
## [1] "Media da coluna SMOTE, FALSE, FALSE = 0.993315446618626"
```

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 = 242.62, 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
##          1.854167          3.720833          3.979167
##  FALSE, TRUE, FALSE  SMOTE, FALSE, FALSE
##          3.750000          1.695833
```

Plotando grafico de Critical Difference

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

