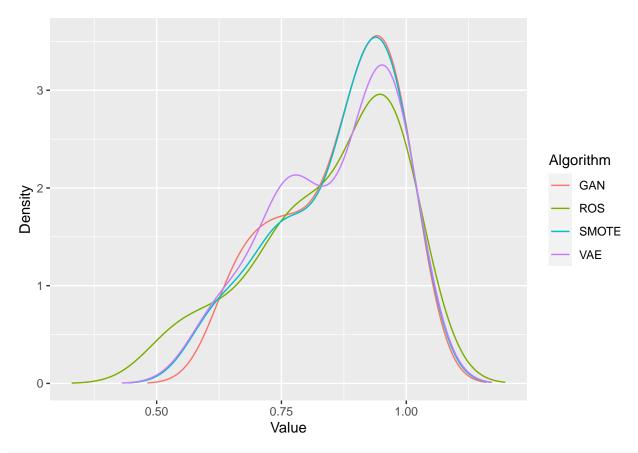
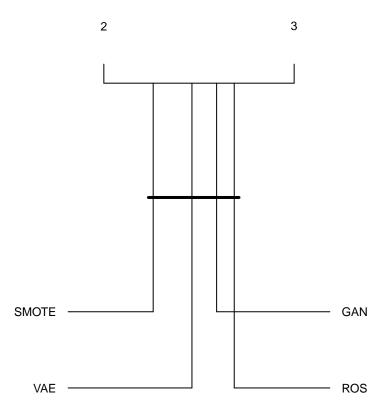
```
library("scmamp")
resultados <- read.csv("test_res_ensemble.csv", header = TRUE, sep = '\t')
tablatst <- cbind(resultados[,1:dim(resultados)[2]])</pre>
colnames(tablatst) <- names(resultados)[1:dim(resultados)[2]]</pre>
colnames(tablatst)
## [1] "SMOTE" "ROS"
                       "VAE"
                               "GAN"
tablatst
##
       SMOTE
                ROS
                       VAE
                              GAN
## 1 0.9105 0.8312 0.8312 0.9182
## 2 0.9821 0.9806 0.9796 0.9800
## 3 0.8942 0.8930 0.8856 0.8853
## 4 1.0000 1.0000 1.0000 0.9976
## 5 0.7653 0.7377 0.7996 0.7718
## 6 0.9330 0.9339 0.9382 0.9313
## 7 0.9331 0.9331 0.9331 0.8729
## 8 0.9607 0.9969 0.9413 0.9782
## 9 0.9589 0.9461 0.9531 0.9537
## 10 0.8574 0.8807 0.8958 0.8723
## 11 0.6142 0.5280 0.6023 0.6535
## 12 0.7295 0.7085 0.6603 0.7039
## 13 0.8944 0.7746 0.8944 0.8944
## 14 0.9965 0.9856 0.9936 0.9821
## 15 0.6108 0.6280 0.6215 0.6416
## 16 0.8944 0.8944 0.7746 0.8944
## 17 0.6639 0.6396 0.7025 0.6722
## 18 0.8511 0.8010 0.7719 0.7800
## 19 1.0000 0.9985 1.0000 1.0000
## 20 0.7916 0.7848 0.7659 0.7401
## 21 0.9879 0.9857 0.9900 0.9945
## 22 0.7849 0.7990 0.7920 0.8369
## 23 0.8986 0.9459 0.9420 0.9383
## 24 0.7409 0.7508 0.7391 0.7299
## 25 0.9401 0.9373 0.9589 0.9400
## 26 0.9650 0.9645 0.9659 0.9595
## 27 0.7226 0.5470 0.7455 0.8022
#Test de Fligner-Killeen para la homocedasticidad (no paramétrico, mediana, no normalidad)
fligner.test(x = list(tablatst$SMOTE, tablatst$ROS, tablatst$VAE, tablatst$GAN))
##
## Fligner-Killeen test of homogeneity of variances
## data: list(tablatst$SMOTE, tablatst$ROS, tablatst$VAE, tablatst$GAN)
## Fligner-Killeen:med chi-squared = 1.3056, df = 3, p-value = 0.7278
#Aplicación del test de Friedman Aligned Ranks
test_friedman_aligned_ranks <- friedmanAlignedRanksTest(as.matrix(tablatst))</pre>
test_friedman_aligned_ranks
##
## Friedman's Aligned Rank Test for Multiple Comparisons
##
```

```
## data: as.matrix(tablatst)
## T = 1.9425, df = 3, p-value = 0.5844
#Aplicación del test de Friedman Aligned Ranks post-hoc
test_friedman_aligned_ranks_post <- friedmanAlignedRanksPost(as.matrix(tablatst))</pre>
test_friedman_aligned_ranks_post
##
             SMOTE
                         ROS
                                   VAE
                                              GAN
## SMOTE
                NA 0.1497833 0.5202055 0.7926587
## ROS
        0.1497833
                          NA 0.4252950 0.2390209
## VAE
         0.5202055 0.4252950
                                    NA 0.7038200
## GAN
         0.7926587 0.2390209 0.7038200
#Bergmann and Hommel dynamic correction of p-values
adjustBergmannHommel (test_friedman_aligned_ranks_post)
##
             SMOTE
                         ROS
                                   VAE
## SMOTE
                NA 0.8986998 1.0000000 1.0000000
## ROS
        0.8986998
                          NA 0.8986998 0.8986998
         1.0000000 0.8986998
                                    NA 1.0000000
## VAE
## GAN
         1.0000000 0.8986998 1.0000000
quadeTest(as.matrix(tablatst))
##
##
   Quade for Multiple Comparisons
##
## data: as.matrix(tablatst)
## T = 0.65056, df = 3, p-value = 0.585
quadePost(as.matrix(tablatst))
             SMOTE
##
                         ROS
                                   VAE
                                              GAN
                NA 0.3540842 0.5455995 0.9785708
## SMOTE
                          NA 0.7472033 0.3403078
         0.3540842
## ROS
         0.5455995 0.7472033
## VAE
                                    NA 0.5278915
## GAN
         0.9785708 0.3403078 0.5278915
                                               NA
plotDensities(as.matrix(tablatst))
```



test <- postHocTest(as.matrix(tablatst), test="friedman", correct="bergmann", use.rank=TRUE)
plotRanking(pvalues=test\$corrected.pval, summary=test\$summary, alpha=0.05)</pre>



```
##TABLA NORMALIZADA - smote (other) vs vae (ref) para WILCOXON
# + 0.1 porque wilcox R falla para valores == 0 en la tabla
difs <- (tablatst[,1] - tablatst[,3]) / tablatst[,1]</pre>
wilc_1_2 <- cbind(ifelse (difs<0, abs(difs)+0.1, 0+0.1), ifelse (difs>0,
                                                                              abs(difs)+0.1, 0+0.1)
colnames(wilc_1_2) <- c(colnames(tablatst)[1], colnames(tablatst)[3])</pre>
head(wilc_1_2)
##
            SMOTE
                        VAE
## [1,] 0.1000000 0.1870950
## [2,] 0.1000000 0.1025456
## [3,] 0.1000000 0.1096175
## [4,] 0.1000000 0.1000000
## [5,] 0.1448190 0.1000000
## [6,] 0.1055734 0.1000000
#Aplicación del test de WILCOXON
SMvsVAEtst <- wilcox.test(wilc_1_2[,1], wilc_1_2[,2], alternative = "two.sided", paired=TRUE)
## Warning in wilcox.test.default(wilc_1_2[, 1], wilc_1_2[, 2], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
Rmas <- SMvsVAEtst$statistic</pre>
pvalue <- SMvsVAEtst$p.value</pre>
SMvsVAEtst <- wilcox.test(wilc_1_2[,2], wilc_1_2[,1], alternative = "two.sided", paired=TRUE)</pre>
## Warning in wilcox.test.default(wilc_1_2[, 2], wilc_1_2[, 1], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
```

```
Rmenos <- SMvsVAEtst$statistic</pre>
Rmas
##
   V
## 123
Rmenos
##
   V
## 153
pvalue
## [1] 0.6592008
##TABLA NORMALIZADA - smote (other) vs gan (ref) para WILCOXON
# + 0.1 porque wilcox R falla para valores == 0 en la tabla
difs <- (tablatst[,1] - tablatst[,4]) / tablatst[,1]</pre>
wilc_1_2 <- cbind(ifelse (difs<0, abs(difs)+0.1, 0+0.1), ifelse (difs>0,
                                                                              abs(difs)+0.1, 0+0.1)
colnames(wilc_1_2) <- c(colnames(tablatst)[1], colnames(tablatst)[4])</pre>
head(wilc_1_2)
##
            SMOTE
                        GAN
## [1,] 0.1084569 0.1000000
## [2,] 0.1000000 0.1021383
## [3,] 0.1000000 0.1099530
## [4,] 0.1000000 0.1024000
## [5,] 0.1084934 0.1000000
## [6,] 0.1000000 0.1018221
#Aplicación del test de WILCOXON
SMvsGANtst <- wilcox.test(wilc_1_2[,1], wilc_1_2[,2], alternative = "two.sided", paired=TRUE)</pre>
## Warning in wilcox.test.default(wilc_1_2[, 1], wilc_1_2[, 2], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
Rmas <- SMvsGANtst$statistic
pvalue <- SMvsGANtst$p.value</pre>
SMvsGANtst <- wilcox.test(wilc_1_2[,2], wilc_1_2[,1], alternative = "two.sided", paired=TRUE)
## Warning in wilcox.test.default(wilc_1_2[, 2], wilc_1_2[, 1], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
Rmenos <- SMvsGANtst$statistic
Rmas
## 164
Rmenos
## 136
pvalue
## [1] 0.6997083
##TABLA NORMALIZADA - vae (other) vs gan (ref) para WILCOXON
# + 0.1 porque wilcox R falla para valores == 0 en la tabla
```

```
difs <- (tablatst[,3] - tablatst[,4]) / tablatst[,3]</pre>
wilc_1_2 <- cbind(ifelse (difs<0, abs(difs)+0.1, 0+0.1), ifelse (difs>0,
                                                                               abs(difs)+0.1, 0+0.1)
colnames(wilc_1_2) <- c(colnames(tablatst)[3], colnames(tablatst)[4])</pre>
head(wilc_1_2)
##
              VAE
## [1,] 0.2046679 0.1000000
## [2,] 0.1004083 0.1000000
## [3,] 0.1000000 0.1003388
## [4,] 0.1000000 0.1024000
## [5,] 0.1000000 0.1347674
## [6,] 0.1000000 0.1073545
#Aplicación del test de WILCOXON
VAEvsGANtst <- wilcox.test(wilc_1_2[,1], wilc_1_2[,2], alternative = "two.sided", paired=TRUE)
## Warning in wilcox.test.default(wilc_1_2[, 1], wilc_1_2[, 2], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
Rmas <- VAEvsGANtst$statistic</pre>
pvalue <- VAEvsGANtst$p.value</pre>
VAEvsGANtst <- wilcox.test(wilc_1_2[,2], wilc_1_2[,1], alternative = "two.sided", paired=TRUE)
## Warning in wilcox.test.default(wilc_1_2[, 2], wilc_1_2[, 1], alternative =
## "two.sided", : cannot compute exact p-value with zeroes
Rmenos <- VAEvsGANtst$statistic</pre>
Rmas
##
     V
## 185
Rmenos
##
     V
## 140
pvalue
## [1] 0.5538827
```