

# Comparison between different local tests: Simes, Simes with Storey and Wilcoxon-Mann-Whitney

27-04-2023

The aim is to compare on real datasets the performance of three closed testing procedures, which respectively use Simes local test with and without Storey estimator for the proportion of true null hypotheses and Wilcoxon-Mann-Whitney local test.

## R functions and libraries

```
library(nout)
library(R.matlab)
```

```
## R.matlab v3.7.0 (2022-08-25 21:52:34 UTC) successfully loaded. See ?R.matlab for help.
```

```
##
## Caricamento pacchetto: 'R.matlab'
```

```
## I seguenti oggetti sono mascherati da 'package:base':
##
##      getOption, isOpen
```

```
library(isotree)
library(farff)
library(tictoc)
```

```
sim_realdata = function(B, dataset, m1, m, n, l, in_index, out_index=NULL, alpha=m/(l+1), lambda = 0.5)
```

```
  m0=m-m1
  if(m1!=0 & is.null(out_index)){
    stop("Error: arg out_index must be initialized.")
  }

  # if(m!=(m1+m0)){
  #   stop("Error: equation m=m1+m0 must be verified.")
  # }
```

```
  if(m1!=0){
    tr_ind = sample(in_index, size = n)
    tr = dataset[tr_ind,]
```

```

iso.fo = isolation.forest(tr, ndim=ncol(dataset), ntrees=10, nthreads=1,
                          scoring_metric = "depth", output_score = TRUE)
in_index2 = setdiff(in_index, tr_ind)

crit=critWMW(m=m, n=n, alpha=alpha)

d_WMW = rep(0,B)
d_Simes = rep(0,B)
d_StoSimes = rep(0,B)
d_BH = rep(0,B)
d_StoBH = rep(0,B)

for(b in 1:B){
  cal_ind = sample(in_index2, size = 1)
  in_index3 = setdiff(in_index2, cal_ind)
  tein_ind = sample(in_index3, size = m0)
  teout_ind = sample(out_index, size = m1)

  cal = dataset[cal_ind,]
  te = dataset[c(tein_ind, teout_ind),]

  S_cal = predict.isolation_forest(iso.fo$model, cal, type = "score")
  S_te = predict.isolation_forest(iso.fo$model, te, type = "score")

  d_WMW[b] = d_mannwhitney(S_X=S_cal, S_Y=S_te, crit=crit)
  d_Simes[b] = d_Simes(S_X=S_cal, S_Y=S_te, alpha=alpha)
  d_StoSimes[b] = d_StoreySimes(S_X=S_cal, S_Y=S_te, alpha=alpha)
  d_BH[b] = d_benjhoch(S_X=S_cal, S_Y=S_te, alpha=alpha)
  d_StoBH[b] = d_StoreyBH(S_X=S_cal, S_Y=S_te, alpha=alpha)
}
}

else{
  tr_ind = sample(in_index, size = n)
  tr = dataset[tr_ind,]
  iso.fo = isolation.forest(tr, ndim=ncol(dataset), ntrees=10, nthreads=1,
                          scoring_metric = "depth", output_score = TRUE)
  in_index2 = setdiff(in_index, tr_ind)

  crit=critWMW(m=m, n=n, alpha=alpha)

  d_WMW = rep(0,B)
  d_Simes = rep(0,B)
  d_StoSimes = rep(0,B)
  d_BH = rep(0,B)
  d_StoBH = rep(0,B)

  for(b in 1:B){
    cal_ind = sample(in_index2, size = 1)
    in_index3 = setdiff(in_index2, cal_ind)
    te_ind = sample(in_index3, size = m0)

    cal = dataset[cal_ind,]

```

```

te = dataset[te_ind,]

S_cal = predict.isolation_forest(iso.fo$model, cal, type = "score")
S_te = predict.isolation_forest(iso.fo$model, te, type = "score")

d_WMW[b] = d_mannwhitney(S_X=S_cal, S_Y=S_te, crit=crit)
d_Simes[b] = d_Simes(S_X=S_cal, S_Y=S_te, alpha=alpha)
d_StoSimes[b] = d_StoreySimes(S_X=S_cal, S_Y=S_te, alpha=alpha)
d_BH[b] = d_benjhoch(S_X=S_cal, S_Y=S_te, alpha=alpha)
d_StoBH[b] = d_StoreyBH(S_X=S_cal, S_Y=S_te, alpha=alpha)
}
}

discov = as.data.frame(cbind("d_BH"=d_BH, "d_StoBH"=d_StoBH, "d_Simes"=d_Simes,
                             "d_StoSimes"=d_StoSimes, "d_WMW"=d_WMW))
colnames(discov) = c("BH", "BHSto", "CTSim", "CTSimSto", "CTWMW")
mean.discov = apply(discov, MARGIN = 2, FUN = mean)

powerGlobalNull = as.data.frame(cbind("d_BH"=d_BH>0, "d_StoBH"=d_StoBH>0, "d_Simes"=d_Simes>0,
                                       "d_StoSimes"=d_StoSimes>0, "d_WMW"=d_WMW>0))
colnames(powerGlobalNull) = c("BH", "BHSto", "CTSim", "CTSimSto", "CTWMW")
mean.powerGlobalNull = apply(powerGlobalNull, MARGIN = 2, FUN = mean)

return(list("discoveries"=discov, "mean.discoveries" = mean.discov,
           "powerGlobalNull"=powerGlobalNull, "mean.powerGlobalNull"=mean.powerGlobalNull,
           "m1"=m1, "alpha"=alpha))
}

```

In the following we set the calibration set and the test set size, respectively  $l$  and  $m$ , so that the nominal level  $\alpha$  is proportional to  $\frac{m}{l+1}$ . The train set size is equal to  $n$  and the number of iterations is  $B = 10^3$ .

## Statlog (Shuttle) dataset

The dataset is available at <http://odds.cs.stonybrook.edu/shuttle-dataset>

```

set.seed(321)

# Initializing parameters
B=10^3
n = 199
l = 199
m = 20
alpha = m/(l+1)
m1s = seq(from=0, to=m, by=1)

data = readMat("G:\\Il mio Drive\\PHD\\Progetto di ricerca\\Conformal Inference Project\\Simulazioni\\7
dataset = cbind(data$X, data$y); colnames(dataset)[ncol(dataset)] = "y"
in_ind = which(dataset[,ncol(dataset)]==0)
out_ind = which(dataset[,ncol(dataset)]==1)

tic()

```

```
res = lapply(m1s,
             function(m1) sim_realdData(B=B, in_index=in_ind, out_index=out_ind,
                                         dataset=dataset,
                                         alpha=alpha,l=l, n=n, m=m, m1=m1))
toc()
```

```
## 194.28 sec elapsed
```

```
# Storing results
store_res = list("mean.discov" = matrix(nrow=length(m1s), ncol = 5),
                 "mean.powerGlobalNull" = matrix(nrow=length(m1s), ncol = 5))
row.names = rep(NA, times=length(m1s))
for(i in 1:length(m1s)){
  row.names[i] = paste("m1 =", m1s[i])
}
rownames(store_res$mean.discov) = row.names
colnames(store_res$mean.discov) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")
rownames(store_res$mean.powerGlobalNull) = row.names
colnames(store_res$mean.powerGlobalNull) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")

for(i in 1:length(res)){
  store_res$mean.discov[i,] = res[[i]]$mean.discov
  store_res$mean.powerGlobalNull[i,] = res[[i]]$mean.powerGlobalNull
}

store_res$mean.discov
```

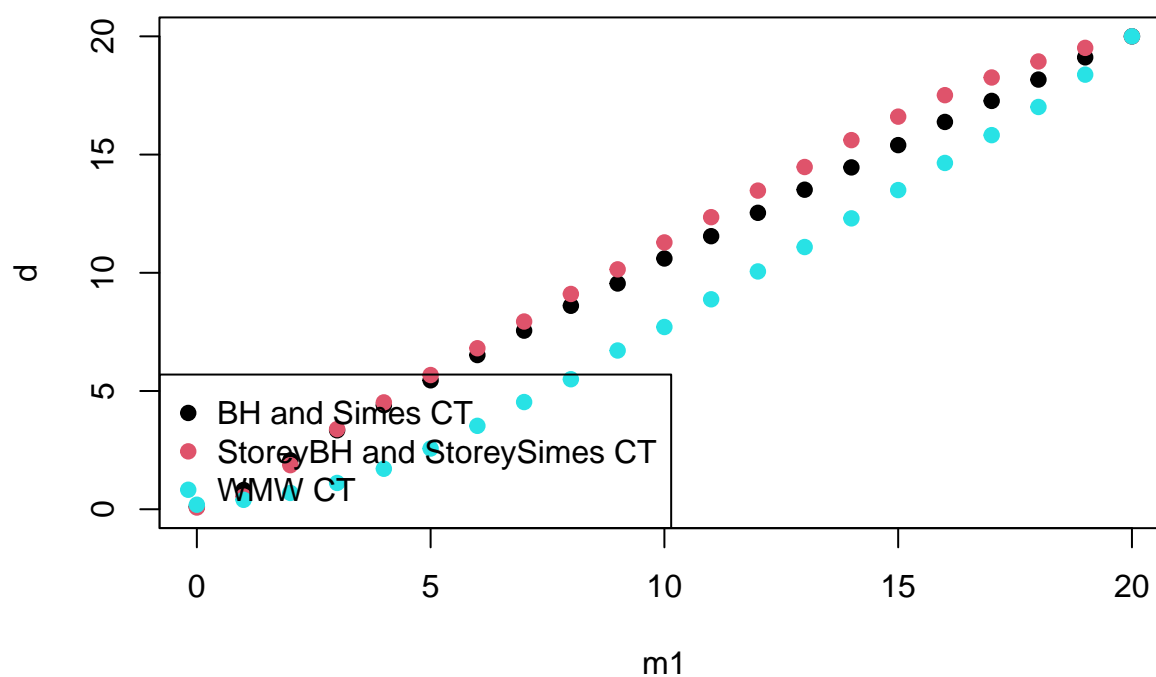
```
##          BH  StoBH  Simes  StoSimes  WMW
## m1 = 0    0.105  0.073  0.085    0.050  0.190
## m1 = 1    0.815  0.536  0.692    0.415  0.398
## m1 = 2    2.068  1.864  1.488    1.172  0.690
## m1 = 3    3.339  3.390  2.891    2.574  1.112
## m1 = 4    4.402  4.519  3.358    3.168  1.712
## m1 = 5    5.453  5.676  4.417    4.336  2.568
## m1 = 6    6.521  6.809  5.355    5.364  3.527
## m1 = 7    7.554  7.940  6.657    6.687  4.534
## m1 = 8    8.605  9.104  7.199    7.289  5.499
## m1 = 9    9.548 10.146  8.296    8.346  6.713
## m1 = 10   10.604 11.285  8.897    9.007  7.707
## m1 = 11   11.549 12.353 10.821   10.665  8.879
## m1 = 12   12.537 13.476 11.885   11.760 10.058
## m1 = 13   13.514 14.474 12.625   12.579 11.089
## m1 = 14   14.457 15.611 13.692   13.703 12.302
## m1 = 15   15.400 16.604 15.043   14.950 13.499
## m1 = 16   16.380 17.512 16.063   15.924 14.646
## m1 = 17   17.270 18.260 17.099   17.005 15.822
## m1 = 18   18.172 18.938 18.090   18.015 17.013
## m1 = 19   19.112 19.513 19.112   19.057 18.381
## m1 = 20   20.000 20.000 20.000   20.000 20.000
```

```
store_res$mean.powerGlobalNull
```

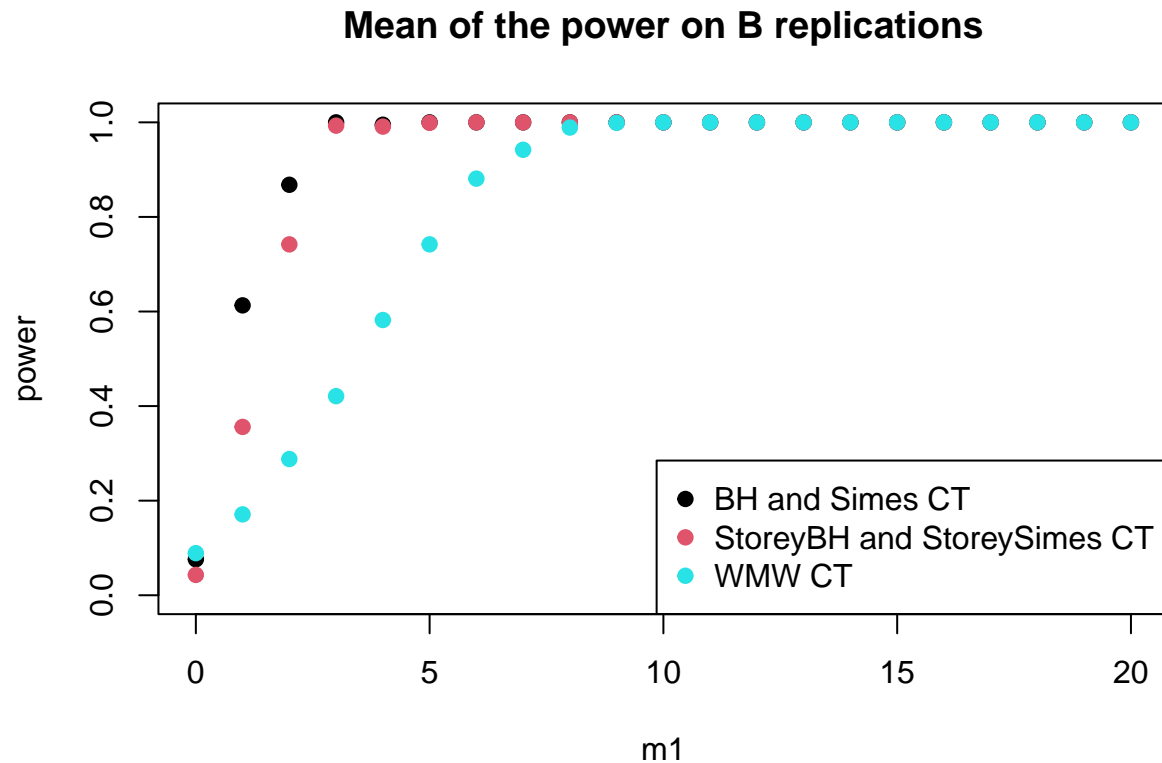
```
##           BH StoBH Simes StoSimes    WMW
## m1 = 0  0.076 0.043 0.076    0.043 0.089
## m1 = 1  0.613 0.356 0.613    0.356 0.171
## m1 = 2  0.868 0.742 0.868    0.742 0.288
## m1 = 3  1.000 0.993 1.000    0.993 0.421
## m1 = 4  0.995 0.991 0.995    0.991 0.582
## m1 = 5  1.000 0.999 1.000    0.999 0.742
## m1 = 6  1.000 1.000 1.000    1.000 0.881
## m1 = 7  1.000 1.000 1.000    1.000 0.942
## m1 = 8  1.000 1.000 1.000    1.000 0.989
## m1 = 9  1.000 1.000 1.000    1.000 0.999
## m1 = 10 1.000 1.000 1.000    1.000 1.000
## m1 = 11 1.000 1.000 1.000    1.000 1.000
## m1 = 12 1.000 1.000 1.000    1.000 1.000
## m1 = 13 1.000 1.000 1.000    1.000 1.000
## m1 = 14 1.000 1.000 1.000    1.000 1.000
## m1 = 15 1.000 1.000 1.000    1.000 1.000
## m1 = 16 1.000 1.000 1.000    1.000 1.000
## m1 = 17 1.000 1.000 1.000    1.000 1.000
## m1 = 18 1.000 1.000 1.000    1.000 1.000
## m1 = 19 1.000 1.000 1.000    1.000 1.000
## m1 = 20 1.000 1.000 1.000    1.000 1.000
```

```
plot(x = m1s, y = store_res$mean.discov[,1], col = 1, ylab = "d",
     xlab = "m1", ylim=c(0,m), pch=19,
     main = "Mean of the number of discoveries on B replications")
points(x = m1s, y = store_res$mean.discov[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.discov[,5], col = 5, pch=19)
legend("bottomleft", pch = 19, col = c(1,2,5),
     legend =c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

## Mean of the number of discoveries on B replications



```
plot(x = m1s, y = store_res$mean.powerGlobalNull[,1], col = 1, ylab = "power",
     xlab = "m1", ylim=c(0,1), pch = 19,
     main = "Mean of the power on B replications")
points(x = m1s, y = store_res$mean.powerGlobalNull[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.powerGlobalNull[,5], col = 5, pch=19)
legend("bottomright", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```



## Credit Card Fraud Detection dataset

The dataset is available at <https://www.kaggle.com/mlg-ulb/creditcardfraud>.

```
set.seed(321)

# Initializing parameters
B=10^3
n = 199
l = 199
m = 20
alpha = m/(l+1)
m1s = seq(from=0, to=m, by=1)

dataset = read.csv("G:\\Il mio Drive\\PHD\\Progetto di ricerca\\Conformal Inference Project\\Simulazioni\\Credit Card Fraud Detection\\dataset.csv")
in_ind = which(dataset[,ncol(dataset)]==0)
out_ind = which(dataset[,ncol(dataset)]==1)

tic()
res = lapply(m1s,
             function(m1) sim_realddata(B=B, in_index=in_ind, out_index=out_ind,
                                         dataset=dataset,
                                         alpha=alpha, l=l, n=n, m=m, m1=m1))
toc()
```

## 492.39 sec elapsed

```
# Storing results
store_res = list("mean.discov" = matrix(nrow=length(m1s), ncol = 5),
                 "mean.powerGlobalNull" = matrix(nrow=length(m1s), ncol = 5))
row.names = rep(NA, times=length(m1s))
for(i in 1:length(m1s)){
  row.names[i] = paste("m1 =", m1s[i])
}
rownames(store_res$mean.discov) = row.names
colnames(store_res$mean.discov) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")
rownames(store_res$mean.powerGlobalNull) = row.names
colnames(store_res$mean.powerGlobalNull) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")

for(i in 1:length(res)){
  store_res$mean.discov[i,] = res[[i]]$mean.discov
  store_res$mean.powerGlobalNull[i,] = res[[i]]$mean.powerGlobalNull
}

store_res$mean.discov
```

##		BH	StoBH	Simes	StoSimes	WMW
##	m1 = 0	0.131	0.100	0.113	0.075	0.215
##	m1 = 1	0.190	0.165	0.140	0.103	0.351
##	m1 = 2	0.733	0.620	0.503	0.371	0.588
##	m1 = 3	1.576	1.547	1.032	0.893	1.043
##	m1 = 4	1.882	1.966	1.078	1.041	1.402
##	m1 = 5	1.413	1.928	0.752	0.876	2.170
##	m1 = 6	3.831	4.357	2.399	2.481	2.908
##	m1 = 7	3.173	4.176	1.641	1.987	3.504
##	m1 = 8	6.337	7.212	3.884	4.251	4.608
##	m1 = 9	7.047	8.486	3.632	4.514	5.742
##	m1 = 10	6.277	8.419	3.017	4.139	6.363
##	m1 = 11	9.034	10.338	5.686	6.406	7.314
##	m1 = 12	8.711	11.816	4.104	6.154	8.762
##	m1 = 13	11.387	12.929	8.404	9.042	9.684
##	m1 = 14	12.243	14.075	8.547	9.564	10.734
##	m1 = 15	13.425	14.966	9.871	10.817	11.578
##	m1 = 16	13.918	16.100	9.861	11.489	12.839
##	m1 = 17	13.286	16.533	7.951	11.039	13.248
##	m1 = 18	15.535	17.990	11.420	13.535	15.166
##	m1 = 19	16.600	18.381	13.155	14.745	15.804
##	m1 = 20	17.489	19.225	14.266	16.227	17.443

```
store_res$mean.powerGlobalNull
```

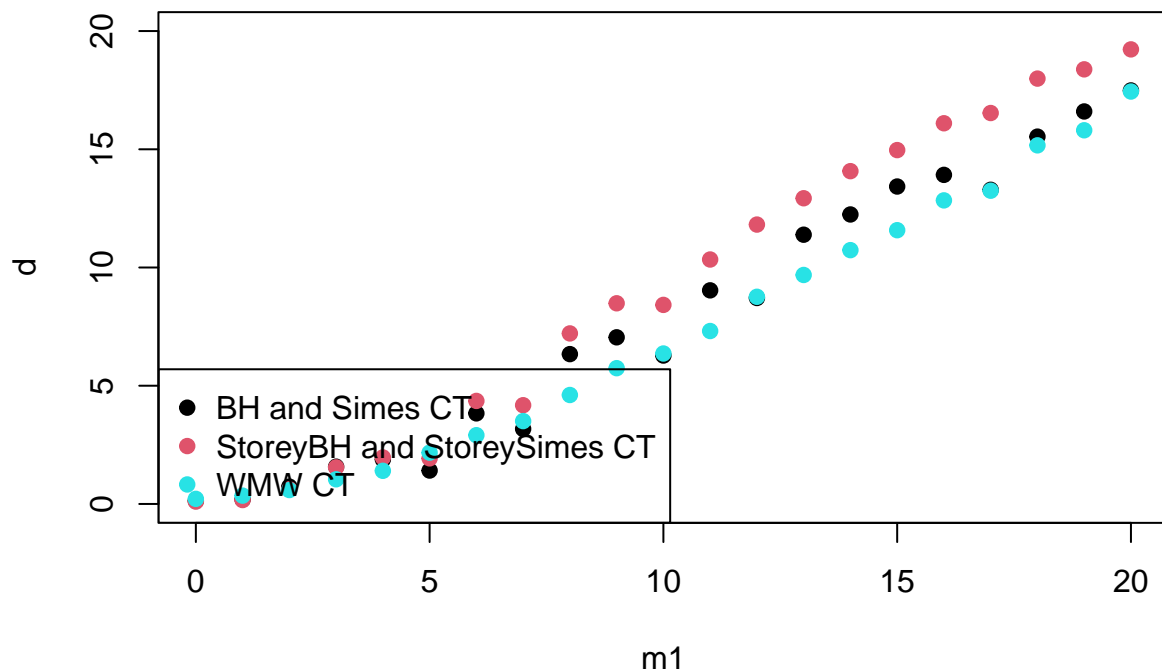
##		BH	StoBH	Simes	StoSimes	WMW
##	m1 = 0	0.094	0.061	0.094	0.061	0.096
##	m1 = 1	0.117	0.085	0.117	0.085	0.166
##	m1 = 2	0.390	0.277	0.390	0.277	0.265
##	m1 = 3	0.649	0.553	0.649	0.553	0.412
##	m1 = 4	0.633	0.580	0.633	0.580	0.507



```
## m1 = 5  0.486 0.504 0.486    0.504 0.650
## m1 = 6  0.938 0.926 0.938    0.926 0.788
## m1 = 7  0.752 0.786 0.752    0.786 0.850
## m1 = 8  0.982 0.987 0.982    0.987 0.943
## m1 = 9  0.948 0.980 0.948    0.980 0.984
## m1 = 10 0.874 0.952 0.874    0.952 0.991
## m1 = 11 0.990 0.997 0.990    0.997 0.999
## m1 = 12 0.940 0.987 0.940    0.987 1.000
## m1 = 13 1.000 1.000 1.000    1.000 1.000
## m1 = 14 0.999 1.000 0.999    1.000 1.000
## m1 = 15 0.997 1.000 0.997    1.000 1.000
## m1 = 16 0.998 1.000 0.998    1.000 1.000
## m1 = 17 0.967 1.000 0.967    1.000 1.000
## m1 = 18 1.000 1.000 1.000    1.000 1.000
## m1 = 19 1.000 1.000 1.000    1.000 1.000
## m1 = 20 1.000 1.000 1.000    1.000 1.000
```

```
plot(x = m1s, y = store_res$mean.discov[,1], col = 1, ylab = "d",
     xlab = "m1", ylim=c(0,m), pch=19,
     main = "Mean of the number of discoveries on B replications")
points(x = m1s, y = store_res$mean.discov[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.discov[,5], col = 5, pch=19)
legend("bottomleft", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

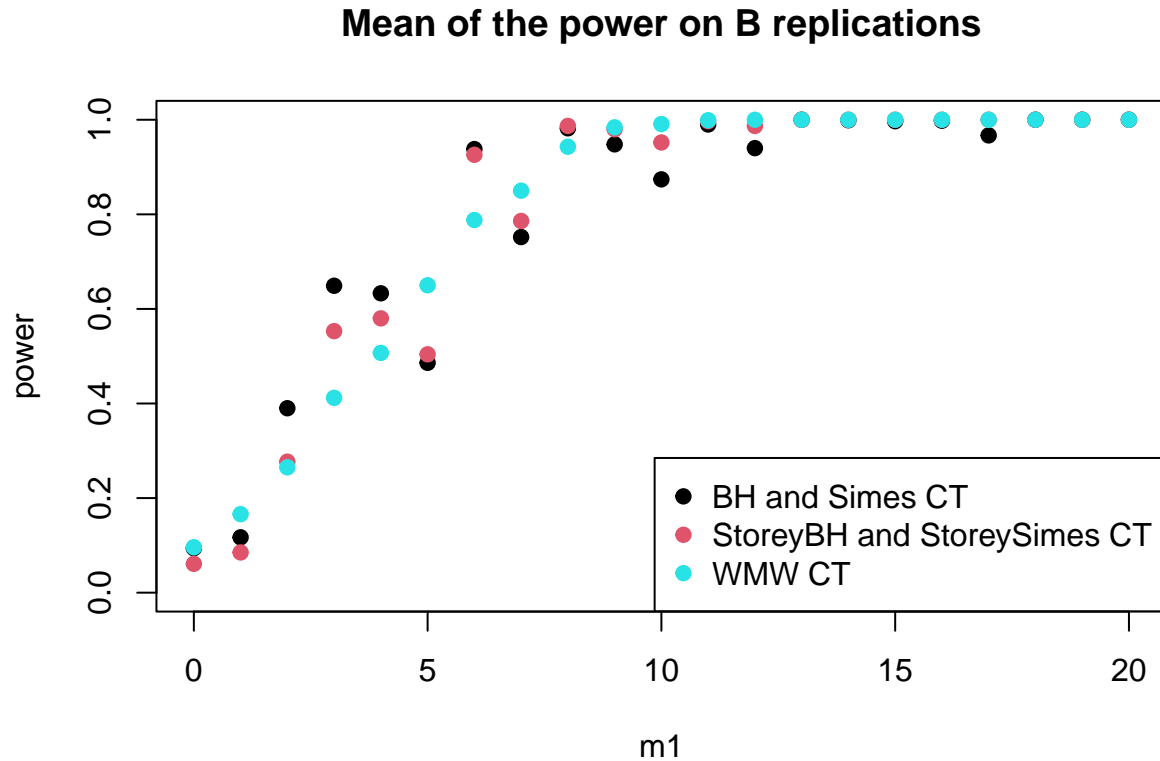
### Mean of the number of discoveries on B replications



```

plot(x = m1s, y = store_res$mean.powerGlobalNull[,1], col = 1, ylab = "power",
     xlab = "m1", ylim=c(0,1), pch = 19,
     main = "Mean of the power on B replications")
points(x = m1s, y = store_res$mean.powerGlobalNull[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.powerGlobalNull[,5], col = 5, pch=19)
legend("bottomright", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))

```



## Covertypes dataset

The dataset is available at <http://odds.cs.stonybrook.edu/forestcovercovertypes-dataset>.

```

set.seed(321)

# Initializing parameters
B=10^3
n = 199
l = 199
m = 20
alpha = m/(l+1)
m1s = seq(from=0, to=m, by=1)

data = readMat("G:\\Il mio Drive\\PHD\\Progetto di ricerca\\Conformal Inference Project\\Simulazioni\\7
dataset = cbind(data$X, data$y); colnames(dataset)[ncol(dataset)] = "y"

```

```

in_ind = which(dataset[,ncol(dataset)]==0)
out_ind = which(dataset[,ncol(dataset)]==1)

tic()
res = lapply(m1s,
             function(m1) sim_realdData(B=B, in_index=in_ind, out_index=out_ind,
                                         dataset=dataset,
                                         alpha=alpha,l=1, n=n, m=m, m1=m1))
toc()

```

```
## 186.92 sec elapsed
```

```

# Storing results
store_res = list("mean.discov" = matrix(nrow=length(m1s), ncol = 5),
                 "mean.powerGlobalNull" = matrix(nrow=length(m1s), ncol = 5))
row.names = rep(NA, times=length(m1s))
for(i in 1:length(m1s)){
  row.names[i] = paste("theta =",m1s[i])
}
rownames(store_res$mean.discov) = row.names
colnames(store_res$mean.discov) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")
rownames(store_res$mean.powerGlobalNull) = row.names
colnames(store_res$mean.powerGlobalNull) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")

for(i in 1:length(res)){
  store_res$mean.discov[i,] = res[[i]]$mean.discov
  store_res$mean.powerGlobalNull[i,] = res[[i]]$mean.powerGlobalNull
}

store_res$mean.discov

```

```

##           BH StoBH Simes StoSimes WMW
## theta = 0  0.105 0.073 0.085  0.050 0.190
## theta = 1  0.815 0.536 0.692  0.415 0.398
## theta = 2  2.068 1.864 1.488  1.172 0.690
## theta = 3  3.339 3.390 2.891  2.574 1.112
## theta = 4  4.402 4.519 3.358  3.168 1.712
## theta = 5  5.453 5.676 4.417  4.336 2.568
## theta = 6  6.521 6.809 5.355  5.364 3.527
## theta = 7  7.554 7.940 6.657  6.687 4.534
## theta = 8  8.605 9.104 7.199  7.289 5.499
## theta = 9  9.548 10.146 8.296  8.346 6.713
## theta = 10 10.604 11.285 8.897  9.007 7.707
## theta = 11 11.549 12.353 10.821 10.665 8.879
## theta = 12 12.537 13.476 11.885 11.760 10.058
## theta = 13 13.514 14.474 12.625 12.579 11.089
## theta = 14 14.457 15.611 13.692 13.703 12.302
## theta = 15 15.400 16.604 15.043 14.950 13.499
## theta = 16 16.380 17.512 16.063 15.924 14.646
## theta = 17 17.270 18.260 17.099 17.005 15.822
## theta = 18 18.172 18.938 18.090 18.015 17.013
## theta = 19 19.112 19.513 19.112 19.057 18.381

```

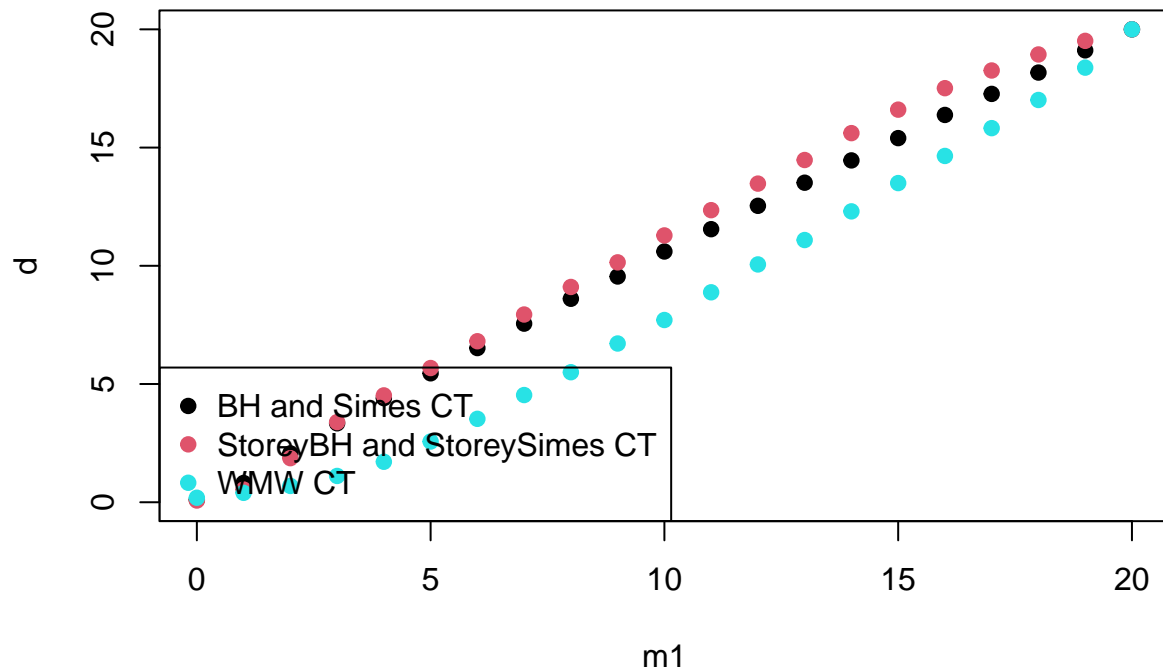
```
## theta = 20 20.000 20.000 20.000 20.000 20.000
```

```
store_res$mean.powerGlobalNull
```

```
##           BH StoBH Simes StoSimes   WMW
## theta = 0  0.076 0.043 0.076    0.043 0.089
## theta = 1  0.613 0.356 0.613    0.356 0.171
## theta = 2  0.868 0.742 0.868    0.742 0.288
## theta = 3  1.000 0.993 1.000    0.993 0.421
## theta = 4  0.995 0.991 0.995    0.991 0.582
## theta = 5  1.000 0.999 1.000    0.999 0.742
## theta = 6  1.000 1.000 1.000    1.000 0.881
## theta = 7  1.000 1.000 1.000    1.000 0.942
## theta = 8  1.000 1.000 1.000    1.000 0.989
## theta = 9  1.000 1.000 1.000    1.000 0.999
## theta = 10 1.000 1.000 1.000    1.000 1.000
## theta = 11 1.000 1.000 1.000    1.000 1.000
## theta = 12 1.000 1.000 1.000    1.000 1.000
## theta = 13 1.000 1.000 1.000    1.000 1.000
## theta = 14 1.000 1.000 1.000    1.000 1.000
## theta = 15 1.000 1.000 1.000    1.000 1.000
## theta = 16 1.000 1.000 1.000    1.000 1.000
## theta = 17 1.000 1.000 1.000    1.000 1.000
## theta = 18 1.000 1.000 1.000    1.000 1.000
## theta = 19 1.000 1.000 1.000    1.000 1.000
## theta = 20 1.000 1.000 1.000    1.000 1.000
```

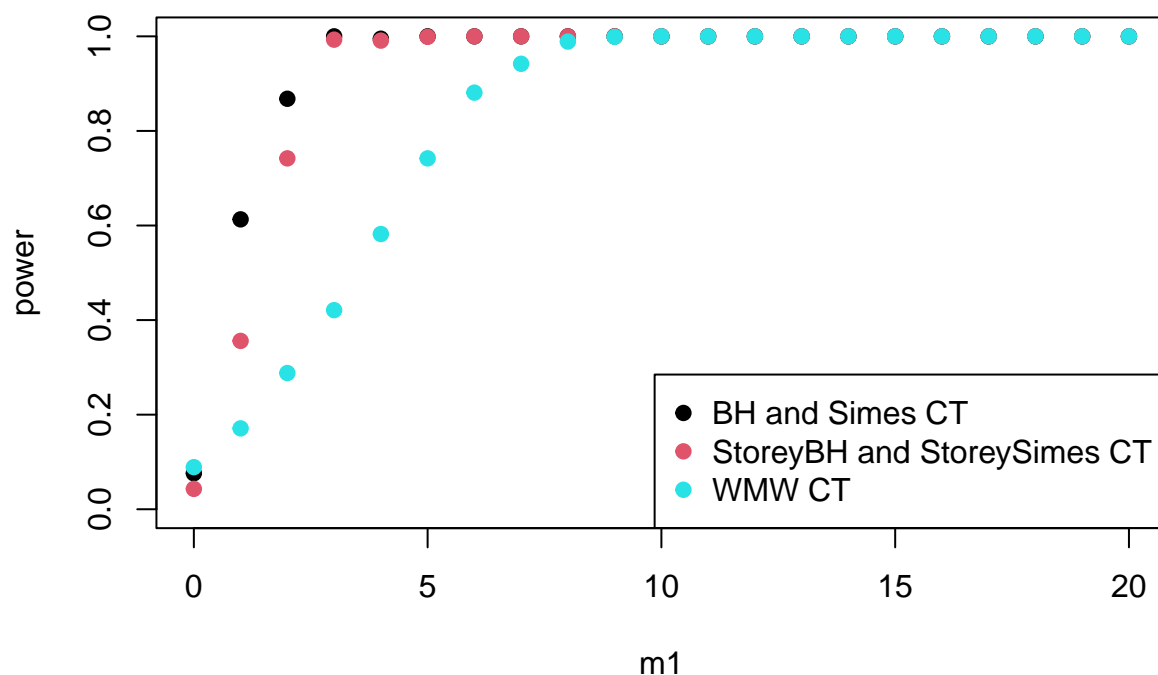
```
plot(x = m1s, y = store_res$mean.discov[,1], col = 1, ylab = "d",
     xlab = "m1", ylim=c(0,m), pch=19,
     main = "Mean of the number of discoveries on B replications")
points(x = m1s, y = store_res$mean.discov[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.discov[,5], col = 5, pch=19)
legend("bottomleft", pch = 19, col = c(1,2,5),
     legend =c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

## Mean of the number of discoveries on B replications



```
plot(x = m1s, y = store_res$mean.powerGlobalNull[,1], col = 1, ylab = "power",
     xlab = "m1", ylim=c(0,1), pch = 19,
     main = "Mean of the power on B replications")
points(x = m1s, y = store_res$mean.powerGlobalNull[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.powerGlobalNull[,5], col = 5, pch=19)
legend("bottomright", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

## Mean of the power on B replications



## Mammography dataset

The dataset is available at <http://odds.cs.stonybrook.edu/mammography-dataset/>.

```
set.seed(321)

# Initializing parameters
B=10^3
n = 199
l = 199
m = 20
alpha = m/(l+1)
m1s = seq(from=0, to=m, by=1)
data = readMat("G:\\Il mio Drive\\PHD\\Progetto di ricerca\\Conformal Inference Project\\Simulazioni\\7
dataset = cbind(data$X, data$y); colnames(dataset)[ncol(dataset)] = "y"
in_ind = which(dataset[,ncol(dataset)]==0)
out_ind = which(dataset[,ncol(dataset)]==1)

tic()
res = lapply(m1s,
  function(m1) sim_realddata(B=B, in_index=in_ind, out_index=out_ind,
    dataset=dataset,
    alpha=alpha,l=l, n=n, m=m, m1=m1))
toc()
```

## 69.7 sec elapsed

```
# Storing results
store_res = list("mean.discov" = matrix(nrow=length(m1s), ncol = 5),
                 "mean.powerGlobalNull" = matrix(nrow=length(m1s), ncol = 5))
row.names = rep(NA, times=length(m1s))
for(i in 1:length(m1s)){
  row.names[i] = paste("theta =", m1s[i])
}
rownames(store_res$mean.discov) = row.names
colnames(store_res$mean.discov) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")
rownames(store_res$mean.powerGlobalNull) = row.names
colnames(store_res$mean.powerGlobalNull) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")

for(i in 1:length(res)){
  store_res$mean.discov[i,] = res[[i]]$mean.discov
  store_res$mean.powerGlobalNull[i,] = res[[i]]$mean.powerGlobalNull
}

store_res$mean.discov
```

##		BH	StoBH	Simes	StoSimes	WMW
##	theta = 0	0.095	0.053	0.078	0.041	0.077
##	theta = 1	0.273	0.208	0.210	0.140	0.152
##	theta = 2	0.396	0.337	0.289	0.210	0.222
##	theta = 3	0.960	0.882	0.676	0.544	0.443
##	theta = 4	0.741	0.727	0.557	0.487	0.557
##	theta = 5	0.562	0.829	0.329	0.393	0.989
##	theta = 6	2.184	2.602	1.126	1.301	1.440
##	theta = 7	1.979	2.627	1.015	1.270	1.898
##	theta = 8	2.667	3.650	1.415	1.749	2.511
##	theta = 9	2.491	3.719	1.469	1.865	3.193
##	theta = 10	3.566	4.946	1.916	2.545	3.575
##	theta = 11	2.394	4.601	1.123	2.060	4.618
##	theta = 12	2.806	6.389	1.405	2.552	5.722
##	theta = 13	5.651	9.040	3.234	4.597	7.194
##	theta = 14	6.275	9.512	3.424	5.107	7.464
##	theta = 15	2.983	6.668	1.567	2.895	6.664
##	theta = 16	7.189	12.123	3.581	6.369	9.385
##	theta = 17	8.092	13.178	4.417	7.171	10.279
##	theta = 18	1.347	12.315	0.431	4.913	10.429
##	theta = 19	4.023	13.641	1.761	6.227	11.184
##	theta = 20	10.842	18.509	6.124	12.046	15.169

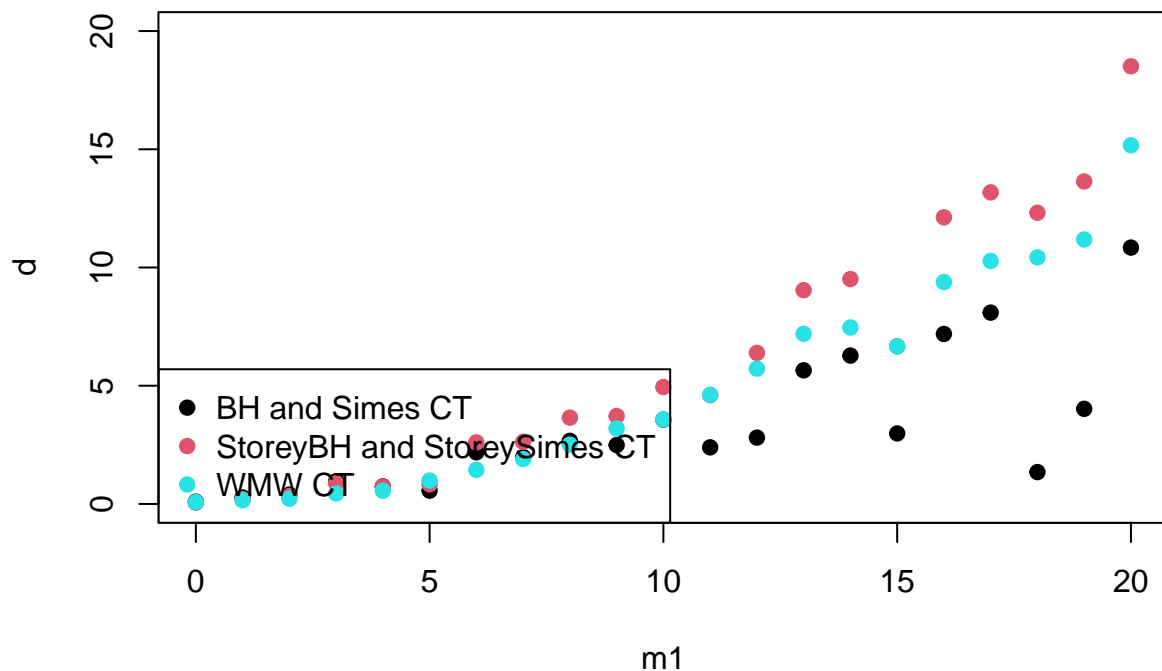
```
store_res$mean.powerGlobalNull
```

##		BH	StoBH	Simes	StoSimes	WMW
##	theta = 0	0.072	0.037	0.072	0.037	0.040
##	theta = 1	0.189	0.126	0.189	0.126	0.074
##	theta = 2	0.249	0.179	0.249	0.179	0.101
##	theta = 3	0.475	0.385	0.475	0.385	0.171
##	theta = 4	0.412	0.349	0.412	0.349	0.234

```
## theta = 5  0.253 0.273 0.253    0.273 0.350
## theta = 6  0.572 0.614 0.572    0.614 0.472
## theta = 7  0.543 0.617 0.543    0.617 0.571
## theta = 8  0.658 0.724 0.658    0.724 0.691
## theta = 9  0.728 0.794 0.728    0.794 0.782
## theta = 10 0.735 0.840 0.735    0.840 0.828
## theta = 11 0.539 0.735 0.539    0.735 0.904
## theta = 12 0.673 0.838 0.673    0.838 0.969
## theta = 13 0.898 0.971 0.898    0.971 0.987
## theta = 14 0.916 0.987 0.916    0.987 0.996
## theta = 15 0.713 0.868 0.713    0.868 0.987
## theta = 16 0.841 0.993 0.841    0.993 0.998
## theta = 17 0.940 0.995 0.940    0.995 0.999
## theta = 18 0.237 0.891 0.237    0.891 1.000
## theta = 19 0.658 0.971 0.658    0.971 1.000
## theta = 20 0.984 1.000 0.984    1.000 1.000
```

```
plot(x = m1s, y = store_res$mean.discov[,1], col = 1, ylab = "d",
     xlab = "m1", ylim=c(0,m), pch=19,
     main = "Mean of the number of discoveries on B replications")
points(x = m1s, y = store_res$mean.discov[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.discov[,5], col = 5, pch=19)
legend("bottomleft", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

### Mean of the number of discoveries on B replications

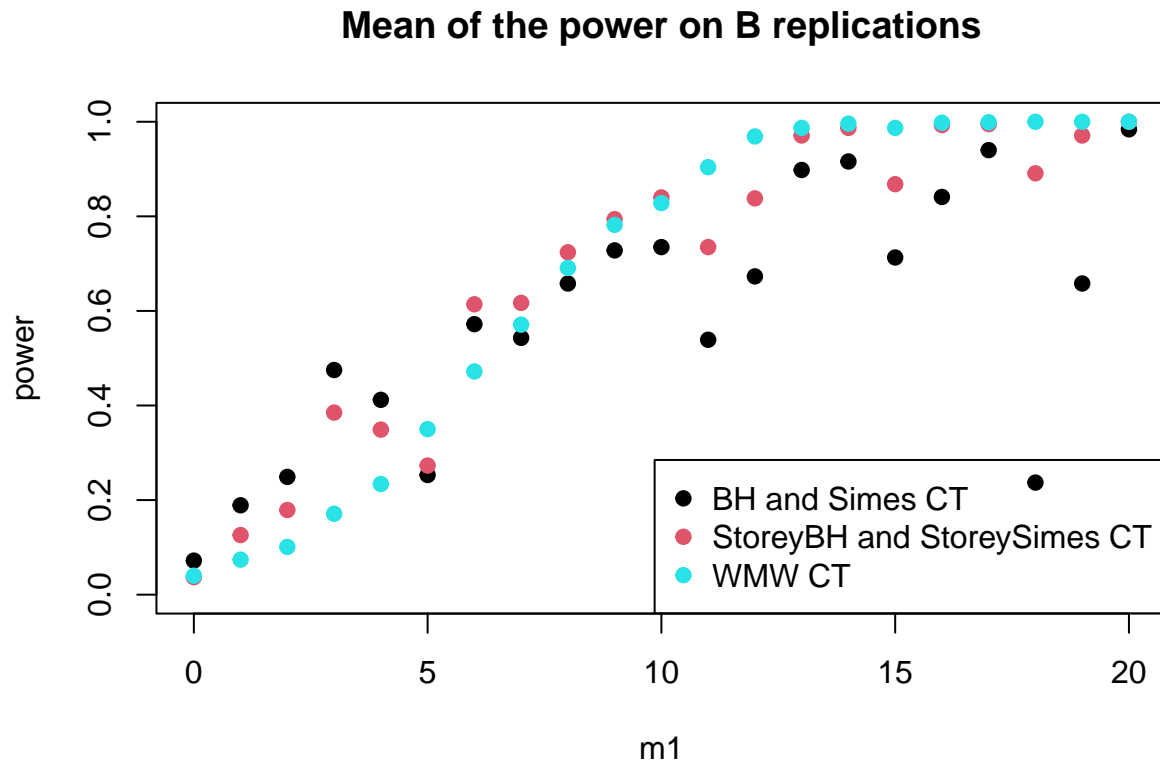




```

plot(x = m1s, y = store_res$mean.powerGlobalNull[,1], col = 1, ylab = "power",
     xlab = "m1", ylim=c(0,1), pch = 19,
     main = "Mean of the power on B replications")
points(x = m1s, y = store_res$mean.powerGlobalNull[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.powerGlobalNull[,5], col = 5, pch=19)
legend("bottomright", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))

```



## Pen-Based Recognition of Handwritten Digits dataset

The dataset is available at <http://odds.cs.stonybrook.edu/pendigits-dataset>.

```

set.seed(321)

# Initializing parameters
B=10^3
n = 199
l = 199
m = 20
alpha = m/(l+1)
m1s = seq(from=0, to=m, by=1)

data = readMat("G:\\Il mio Drive\\PHD\\Progetto di ricerca\\Conformal Inference Project\\Simulazioni\\7
dataset = cbind(data$X, data$y); colnames(dataset)[ncol(dataset)] = "y"

```

```

in_ind = which(dataset[,ncol(dataset)]==0)
out_ind = which(dataset[,ncol(dataset)]==1)

tic()
res = lapply(m1s,
             function(m1) sim_realdData(B=B, in_index=in_ind, out_index=out_ind,
                                         dataset=dataset,
                                         alpha=alpha,l=1, n=n, m=m, m1=m1))
toc()

```

## 98 sec elapsed

```

# Storing results
store_res = list("mean.discov" = matrix(nrow=length(m1s), ncol = 5),
                 "mean.powerGlobalNull" = matrix(nrow=length(m1s), ncol = 5))
row.names = rep(NA, times=length(m1s))
for(i in 1:length(m1s)){
  row.names[i] = paste("theta =",m1s[i])
}
rownames(store_res$mean.discov) = row.names
colnames(store_res$mean.discov) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")
rownames(store_res$mean.powerGlobalNull) = row.names
colnames(store_res$mean.powerGlobalNull) = c("BH", "StoBH", "Simes", "StoSimes", "WMW")

for(i in 1:length(res)){
  store_res$mean.discov[i,] = res[[i]]$mean.discov
  store_res$mean.powerGlobalNull[i,] = res[[i]]$mean.powerGlobalNull
}

store_res$mean.discov

```

##		BH	StoBH	Simes	StoSimes	WMW
##	theta = 0	0.122	0.088	0.098	0.061	0.230
##	theta = 1	0.156	0.126	0.135	0.090	0.335
##	theta = 2	0.344	0.322	0.269	0.197	0.502
##	theta = 3	0.190	0.230	0.135	0.118	0.698
##	theta = 4	0.762	0.891	0.458	0.453	1.362
##	theta = 5	0.431	0.533	0.317	0.329	1.517
##	theta = 6	0.104	0.158	0.086	0.100	1.110
##	theta = 7	1.057	1.697	0.641	0.839	2.931
##	theta = 8	0.779	1.160	0.465	0.589	2.104
##	theta = 9	4.200	4.942	2.441	2.730	3.572
##	theta = 10	2.044	3.330	1.311	1.718	4.812
##	theta = 11	0.251	2.169	0.123	0.669	5.748
##	theta = 12	0.990	3.427	0.529	1.398	6.402
##	theta = 13	0.585	2.080	0.419	0.869	4.466
##	theta = 14	3.156	13.980	1.195	5.756	11.256
##	theta = 15	0.423	9.081	0.264	2.930	10.548
##	theta = 16	2.220	10.634	1.126	4.140	9.438
##	theta = 17	8.564	13.699	4.887	7.705	11.073
##	theta = 18	2.340	5.065	1.465	2.555	5.881
##	theta = 19	10.219	19.505	4.624	14.917	17.476

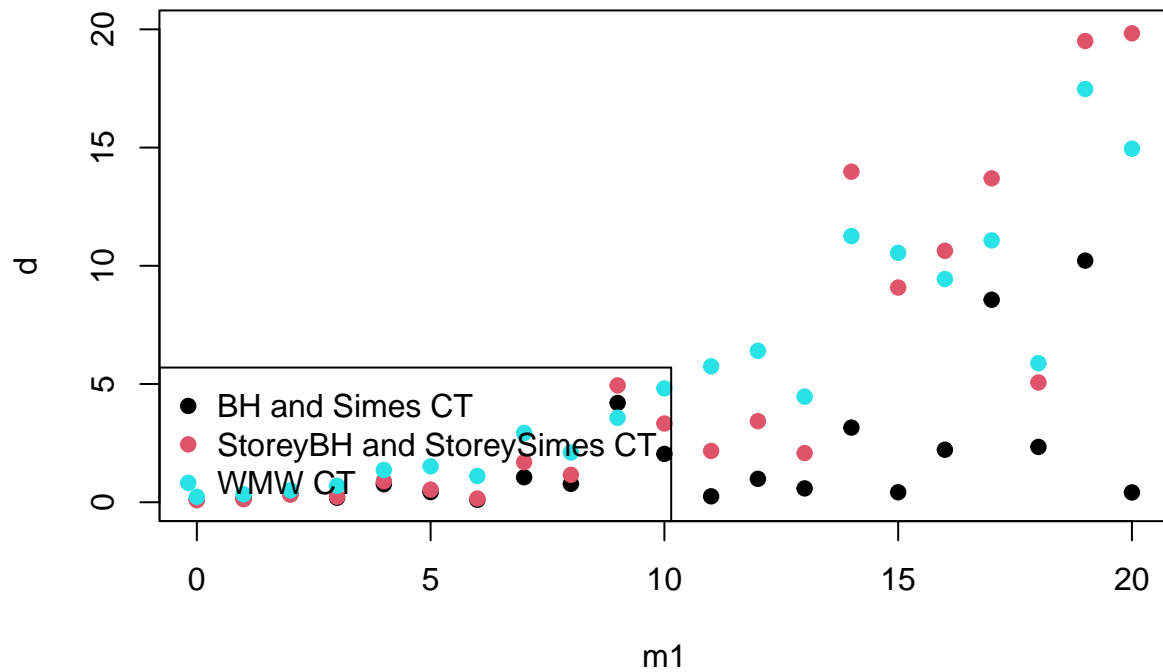
```
## theta = 20 0.416 19.832 0.284 12.291 14.952
```

```
store_res$mean.powerGlobalNull
```

```
##           BH StoBH Simes StoSimes   WMW
## theta = 0 0.095 0.057 0.095    0.057 0.111
## theta = 1 0.120 0.081 0.120    0.081 0.150
## theta = 2 0.222 0.160 0.222    0.160 0.221
## theta = 3 0.117 0.098 0.117    0.098 0.275
## theta = 4 0.322 0.310 0.322    0.310 0.479
## theta = 5 0.253 0.250 0.253    0.250 0.518
## theta = 6 0.082 0.091 0.082    0.091 0.379
## theta = 7 0.428 0.494 0.428    0.494 0.790
## theta = 8 0.325 0.363 0.325    0.363 0.620
## theta = 9 0.879 0.887 0.879    0.887 0.823
## theta = 10 0.747 0.816 0.747    0.816 0.941
## theta = 11 0.089 0.311 0.089    0.311 0.981
## theta = 12 0.344 0.629 0.344    0.629 0.988
## theta = 13 0.329 0.480 0.329    0.480 0.920
## theta = 14 0.522 0.965 0.522    0.965 1.000
## theta = 15 0.216 0.738 0.216    0.738 1.000
## theta = 16 0.611 0.937 0.611    0.937 1.000
## theta = 17 0.972 0.997 0.972    0.997 1.000
## theta = 18 0.758 0.880 0.758    0.880 0.985
## theta = 19 0.859 1.000 0.859    1.000 1.000
## theta = 20 0.223 0.996 0.223    0.996 1.000
```

```
plot(x = m1s, y = store_res$mean.discov[,1], col = 1, ylab = "d",
     xlab = "m1", ylim=c(0,m), pch=19,
     main = "Mean of the number of discoveries on B replications")
points(x = m1s, y = store_res$mean.discov[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.discov[,5], col = 5, pch=19)
legend("bottomleft", pch = 19, col = c(1,2,5),
     legend =c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

## Mean of the number of discoveries on B replications



```
plot(x = m1s, y = store_res$mean.powerGlobalNull[,1], col = 1, ylab = "power",
     xlab = "m1", ylim=c(0,1), pch = 19,
     main = "Mean of the power on B replications")
points(x = m1s, y = store_res$mean.powerGlobalNull[,2], col = 2, pch=19)
points(x = m1s, y = store_res$mean.powerGlobalNull[,5], col = 5, pch=19)
legend("bottomright", pch = 19, col = c(1,2,5),
      legend = c("BH and Simes CT", "StoreyBH and StoreySimes CT", "WMW CT"))
```

Mean of the power on B replications

