# Principles of Statistical Data Analysis: HW3

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### 1 R function: median.test(x,y)

The function median.test(x,y) calculates a permutation p-value associated with  $H_0: F_x = F_y$  versus  $H_A: median_x \neq median_y$ . If the total amount of combinations exceeds 10000, the null-distribution is generated from 10000 random samples. The code is listed below, the included R comments can guide the reader through the code.

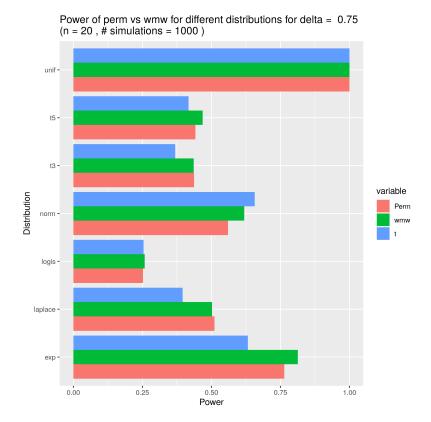
```
# FUNCTION: calc.median.diff: -----
# Function: calculate the difference in median between two groups.
# param: ind = the indices of the first group
# param: vec = the complete list with data from group 1 and group 2
calc.median.diff <- function(ind, vec){</pre>
  # make both groups
  group1 <- vec[ind]</pre>
  group2 <- vec[!(1:length(vec)) %in% ind]</pre>
  # calculate the difference in medians
  return(median(group1) - median(group2))
}
# FUNCTION: median.test: -----
# Function: calculate the p-value of the median difference between
# vector x and vector y, based on the null hypothesis F1(x) = F2(x)
# When the number of combinations is sufficiently small to do a full
# permutation test (limit at 15000), the full permutation test will
# be performed.
```

```
# param: x = vector values for group 1
# param: y = vector values group 2
median.test <- function(x, y){</pre>
  N <- 10000
  realization <- median(x) - median(y)</pre>
  len_x <- length(x)</pre>
  len_y <- length(y)</pre>
  vec \leftarrow c(x, y)
  len_vec <- len_x + len_y</pre>
  if(choose(n = len_x+len_y, len_x) < N){</pre>
    #cat('A limited number of combinations: full permutation test')
    median.diff <- combn(len_vec,</pre>
                           len_x,
                           function(ind){
                             return(calc.median.diff(ind, vec))
                           })
  } else {
    #cat('Too many combinations - A sample test will be performed.\n')
    median.diff <- replicate(N,</pre>
                                calc.median.diff(
                                  sample(c(1:len_vec),len_x),
    )
  }
  if(FALSE){
    hist(abs(median.diff), main = paste('p = ',
                                            mean(abs(realization) <= abs(median.diff))))</pre>
    abline(v = abs(realization), col = "red",
            lty = 1, lwd = 2)
  }
  # The distribution will be symmetrical. The p-value can be calculated by taking
  # the absolute value.
  return(mean(abs(realization) <= abs(median.diff)))</pre>
}
```

# 2 Comparison to other tests

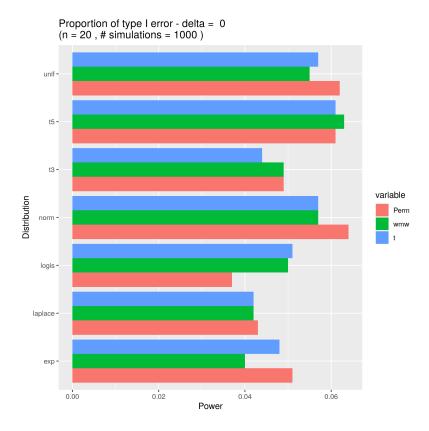
#### 2.1 Power

To evaluate the power of the median test as compared to other statistical tests, the proportions of true positive outcomes of the median test, the permutation t-test and the Wilcoxon–Mann–Whitney test were acquired through Monte Carlo simulation. Samples (n = 20 and n = 40) were randomly drawn from different distributions under  $H_a$  (i.e. for a given delta  $\neq$  0).



## 2.2 Type-I error rate

Next, the simulations were repeated under  $H_0$  (i.e. with delta set equal to zero) to compare the type-I error rate of the median test with the permutation t-test and the Wilcoxon–Mann–Whitney test.



### 3 Conclusion

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### 4 Addendum: full code

```
# HOMEWORK 3
# Group nr 15
# Install and load packages
packages <- c('coin','ggplot2', 'reshape2', 'rmutil')</pre>
# install.packages(packages)
lapply(packages, library, character.only = TRUE)
# FUNCTION: calc.median.diff: -----
# Function: calculate the difference in median between two groups.
# param: ind = the indices of the first group
# param: vec = the complete list with data from group 1 and group 2
calc.median.diff <- function(ind, vec){</pre>
  # make both groups
  group1 <- vec[ind]</pre>
 group2 <- vec[!(1:length(vec)) %in% ind]</pre>
  # calculate the difference in medians
  return(median(group1) - median(group2))
}
# FUNCTION: median.test: -----
\# Function: calculate the p-value of the median difference between
# vector x and vector y, based on the null hypothesis F1(x) = F2(x)
# When the number of combinations is sufficiently small to do a full
# permutation test (limit at 15000), the full permutation test will
# be performed.
# param: x = vector values for group 1
# param: y = vector values group 2
median.test <- function(x, y){</pre>
  N <- 10000
  realization <- median(x) - median(y)
  len_x <- length(x)</pre>
  len_y <- length(y)</pre>
  vec \leftarrow c(x, y)
  len_vec <- len_x + len_y</pre>
  if(choose(n = len_x+len_y, len_x) < N){</pre>
    #cat('A limited number of combinations: full permutation test')
    median.diff <- combn(len_vec,</pre>
                          len x,
                          function(ind){
                            return(calc.median.diff(ind, vec))
```

```
})
  } else {
    \#cat('Too\ many\ combinations - A\ sample\ test\ will\ be\ performed.\n')
    median.diff <- replicate(N,</pre>
                               calc.median.diff(
                                 sample(c(1:len_vec),len_x),
                                 vec)
    )
  }
  if(FALSE){
    hist(abs(median.diff), main = paste('p = ',
                                          mean(abs(realization) <= abs(median.diff))))</pre>
    abline(v = abs(realization), col = "red",
           lty = 1, lwd = 2)
  # The distribution will be symmetrical. The p-value can be calculated by taking
  # the absolute value.
  return(mean(abs(realization) <= abs(median.diff)))</pre>
# Simulations
n <- 20
                 # sample size
N < -200
                 # test simulations
distributions <- c('rt', 'rt', 'rexp', 'rlogis', 'rnorm', 'runif', 'rlaplace')</pre>
dist_names <- c('t3', 't5', 'exp', 'logis', 'norm', 'unif', 'laplace')</pre>
dist_args <- list(list(n, 3), list(n, 5), list(n), list(n), list(n), list(n), list(n))
power.med <- power.wmw <- power.t <- c()</pre>
# For simulations under HO (delta = 0) and Ha (delta = 1)
for(delta in 0:1) {
  if(delta) {
    title <- paste0('Power_', n, '_', N) }</pre>
  else{
    title <- paste0('TypeI_error_', n, '_', N) }</pre>
  # for different distributions
  for(i in 1:length(distributions)) {
    cat(paste('power calculation for distribution : ',
              distributions[i], '\n'))
    p.med <- p.wmw <- p.t <- c()
    # perform N tests on randomly drawn samples Y1 and Y2
    for(j in 1:N) {
      Y1 <- do.call(what = distributions[i],
                     args = dist_args[[i]])
      Y2 <- do.call(what = distributions[i],
                     args = dist_args[[i]]) + delta * sd(Y1)/2
      df \leftarrow data.frame(rep(c('A', 'B'), each = n), c(Y1, Y2))
```

```
colnames(df) <- c("group", "Y")</pre>
      p.med[j] \leftarrow median.test(x = Y1, y = Y2)
      p.wmw[j] <- wilcox.test(Y1, Y2, exact = TRUE)$p.value
      p.t[j] <- pvalue(oneway_test(Y~group, data = df,</pre>
                                      distribution = approximate(nresample = 10000)))
    # calculate the proportion of HO rejections
    power.med[i] <- mean(p.med < 0.05)</pre>
    power.wmw[i] \leftarrow mean(p.wmw \leftarrow 0.05)
    power.t[i] \leftarrow mean(p.t < 0.05)
  df_power <- data.frame(Distribution = dist_names,</pre>
                           med = rep(x = 0.0, length(distributions)),
                           wmw = rep(x = 0.0, length(distributions)),
                           t = rep(x = 0.0, length(distributions)))
  df_power$med <- power.med</pre>
  df_power$wmw <- power.wmw</pre>
  df_power$t <- power.t</pre>
  df_power <- melt(df_power, id.vars = 'Distribution')</pre>
  write.csv(df_power, paste(title, '.csv', sep=''))
  p <- ggplot(data = df_power, aes(x = Distribution, y = value, fill = variable)) +</pre>
    geom_bar(position = 'dodge', stat = 'identity') +
    ggtitle(title) +
    xlab('Distribution') +
    ylab('Power') +
    coord_flip()
  ggsave( paste(title, '.png', sep = ''),
           plot = p,
           device = 'png'
           )
}
```