## Sheet 7

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## Exercise 22

implementation of the bicluster consists a couple of functions first one is the computation of the residuals and  $H_Y$  for a given matrix:

```
# computes the heterogeneity-index for a given matrix m
biclustering residuals <- function(m) {</pre>
  # apparently a matrix with one row gets converted to a vector
  if (is.matrix(m) == T) {
    n_col <- length(m[1,])</pre>
    n_row <- length(m[,1])</pre>
  }else {
    return(NA)
  \#\ basic\ initialization
  overall_mean <- mean(m)
  column_wise_mean <- c()</pre>
  row_wise_mean <- c()</pre>
  residuals <- matrix(data = rep(NA, n_col*n_row),
                                nrow = n_row,
                                ncol = n_col)
  # computing column-wise and row-wise mean
  for (i in 1:n col) {
    column_wise_mean <- append(column_wise_mean, mean(m[,i]) - overall_mean)</pre>
  for (j in 1:n_row) {
    row_wise_mean <- append(row_wise_mean, mean(m[j,]) - overall_mean)</pre>
  # computing SQUARED residuals
  for (i in 1:n_row) {
    for (j in 1:n_col) {
      residuals[i,j] <- (overall_mean + row_wise_mean[i] + column_wise_mean[j] - m[i,j])^2</pre>
  }
  # return sum of residuals
  return(residuals)
h_y <- function(m) {</pre>
```

```
return(sum(biclustering_residuals(m)))
}
```

Next we want to implement the biclustering-algorithm of method 2 for one cluster:

```
bicluster <- function(m, delta) {</pre>
  if (is.matrix(m) == T) {
    n_col <- length(m[1,])</pre>
    n_row <- length(m[,1])</pre>
  }else {
    return(m)
  # check if contributions need to be checked
  if (length(m) == 0) {
    return(m)
  h_y_m \leftarrow h_y(m)
  if (h_y_m \leftarrow delta) {
    return(m)
  # else check contributions
  contr_row <- c()</pre>
  contr_col <- c()</pre>
  bi_res <- biclustering_residuals(m)</pre>
  # determine contributions
  for(i in 1:n_col) {
    contr_col <- append(contr_col, mean(bi_res[,i]))</pre>
  for(j in 1:n_row) {
    contr_row <- append(contr_row, mean(bi_res[j,]))</pre>
  # check which contribution is the biggest
  rm_col <- which.max(contr_col)</pre>
  rm_row <- which.max(contr_row)</pre>
  bgst_contr_r <- contr_row[rm_row]</pre>
  bgst_contr_c <- contr_col[rm_col]</pre>
  # remove the biggest contribution and repeat algorithm
  if (bgst_contr_c >= bgst_contr_r) {
    bicluster(m[,-rm_col], delta)
  } else {
    bicluster(m[-rm_row,], delta)
}
```

Okay so now we can test the implementation for the given data:

```
## [,1] [,2] [,3]
## [1,] 1.03 3.09 7.95
## [2,] 1.99 3.95 9.01
## [3,] 2.98 4.98 10.09
```