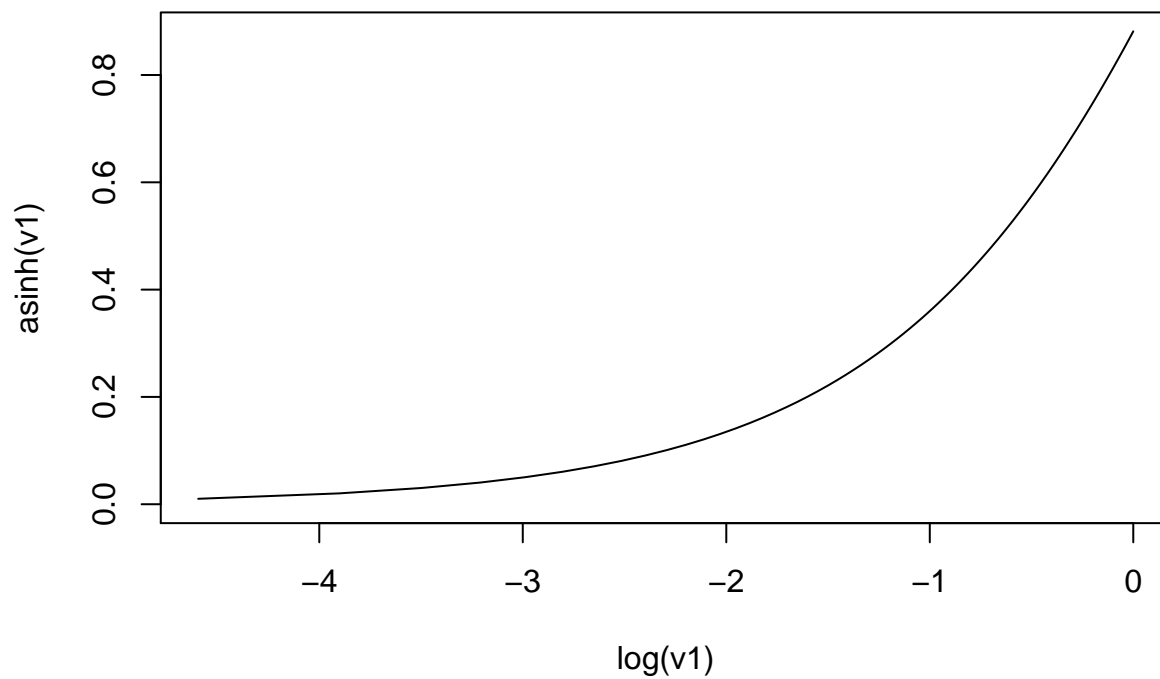


MSMB-Chapter5-Clustering

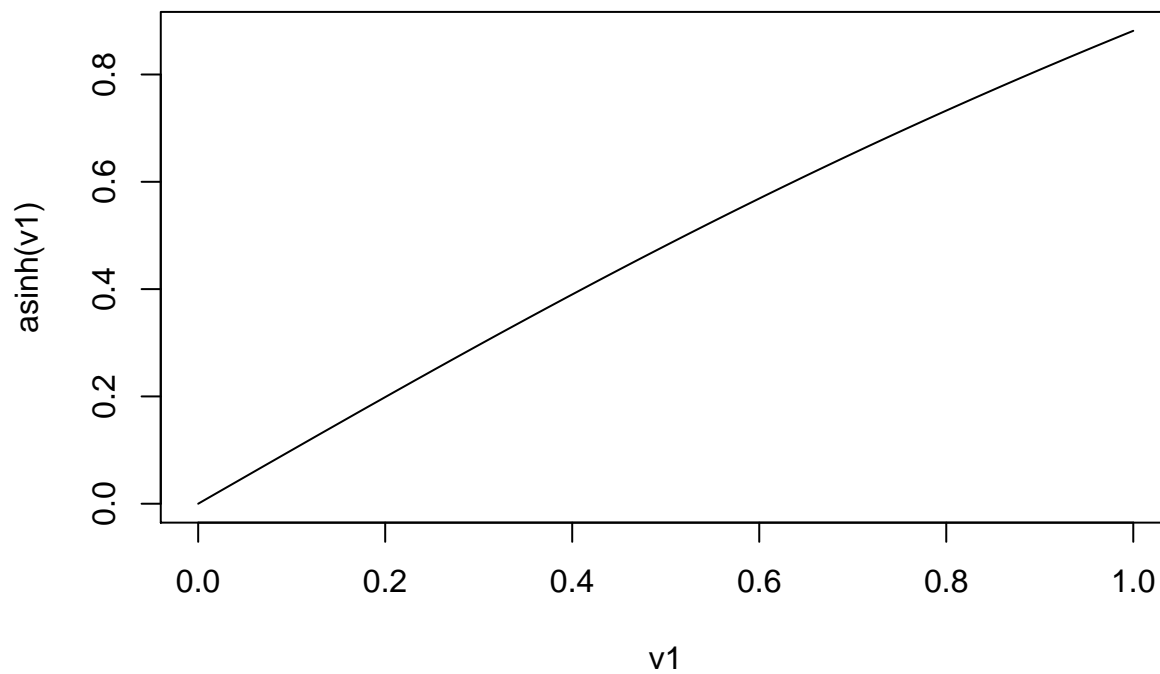
Aleeza Gerstein

2019-11-06

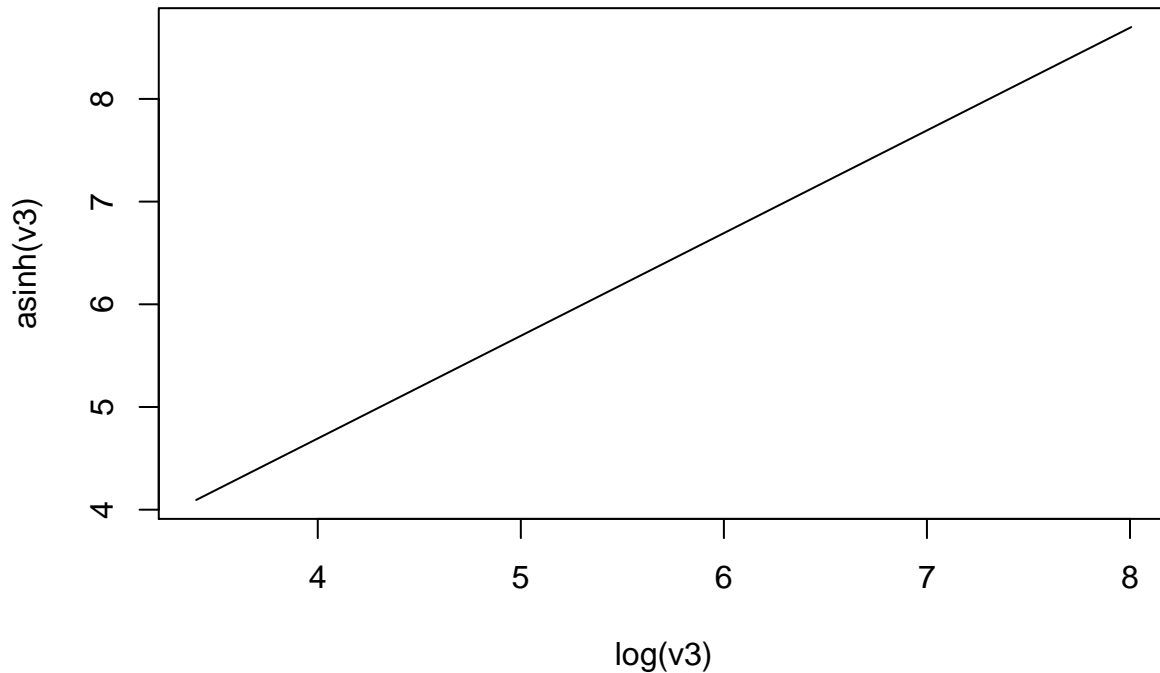
```
v1 <- seq(0, 1, length.out=100)  
plot(log(v1), asinh(v1), type="l")
```



```
plot(v1, asinh(v1), type= "l")
```



```
v3 <- seq(30, 3000, length = 100)
plot(log(v3), asinh(v3), type="l")
```



5.5 Clustering examples: Flow cytometry

```
library("flowCore")
library("flowViz")

## Loading required package: lattice

##
## Attaching package: 'lattice'

## The following objects are masked from 'package:ncdfFlow':
##
##   densityplot, histogram, xyplot

fcsB <- read.FCS("../data/Bendall_2011.fcs")
markersB <- readr::read_csv("../data/Bendall_2011_markers.csv")

## Parsed with column specification:
## cols(
##   isotope = col_character(),
##   marker = col_character()
## )

mt <- match(markersB$isotope, colnames(fcsB))
colnames(fcsB)[mt] <- markersB$marker

asinhtrsrf <- arcsinhTransform(a = 0.1, b = 1)
fcsBT <- transform(fcsB, transformList(colnames(fcsB)[-c(1, 2, 4)], asinhtrsrf))
```

```

kf <- kmeansFilter("CD3all" = c("Pop1", "Pop2"), filterID = "myKmFilter")
fres <- flowCore::filter(fcsBT, kf)
summary(fres)

```

```

## Pop1: 33429 of 91392 events (36.58%)
## Pop2: 57963 of 91392 events (63.42%)

```

```

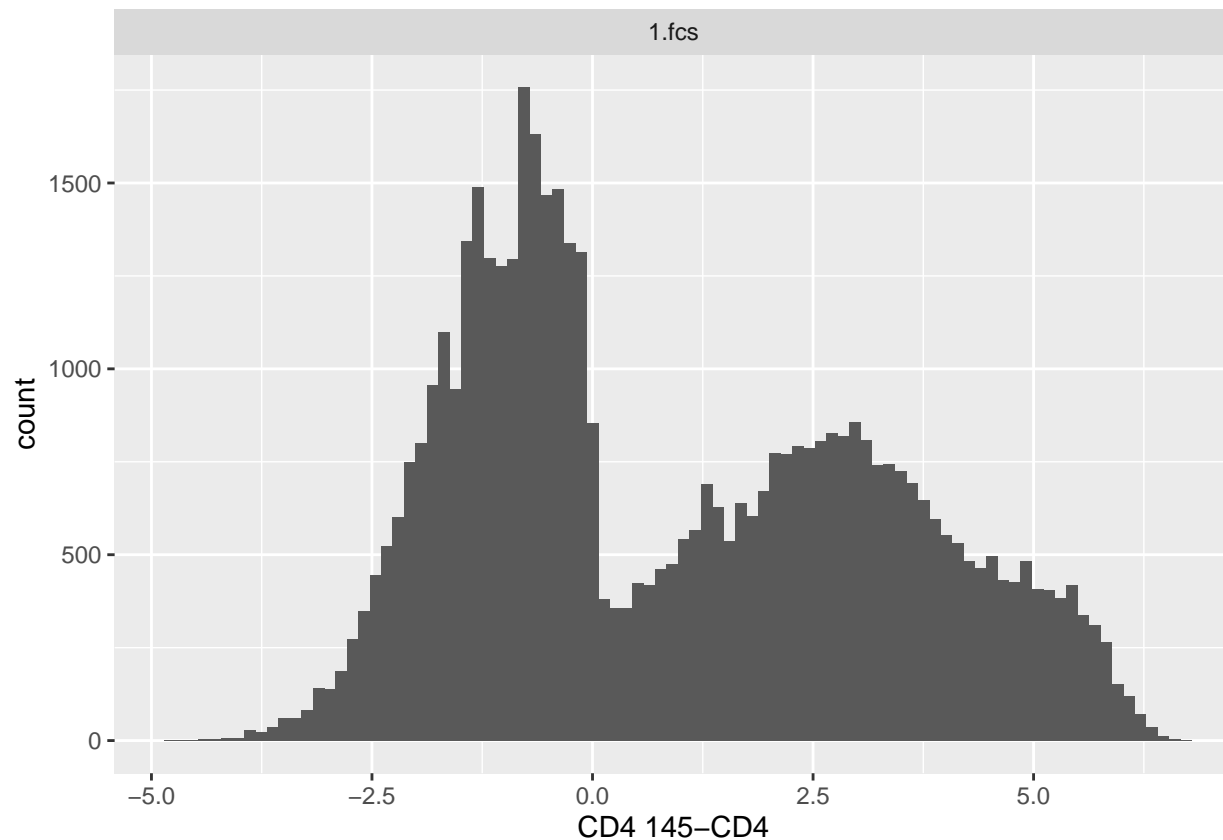
fcsBT1 <- flowCore::split(fcsBT, fres, population = "Pop1")
fcsBT2 <- flowCore::split(fcsBT, fres, population = "Pop2")

```

```

library("ggcyto")
ggcd4cd8 <- ggcyto(fcsB, aes(x = CD4, y = CD8))
ggcd4 <- ggcyto(fcsB, aes(x = CD4))
ggcd8 <- ggcyto(fcsB, aes(x = CD8))
p1 <- ggcd4 +
  geom_histogram(bins = 60)
p1b <- ggcd8 +
  geom_histogram(bins = 60)
asinht <- arcsinhTransform(a = 0, b = 1)
trans1 <- transformList(colnames(fcsB)[-c(1, 2, 4)], asinht)
fcsBT <- transform(fcsB, trans1)
p1t <- ggcyto(fcsBT, aes(x=CD4)) +
  geom_histogram(bins = 90)
p1t

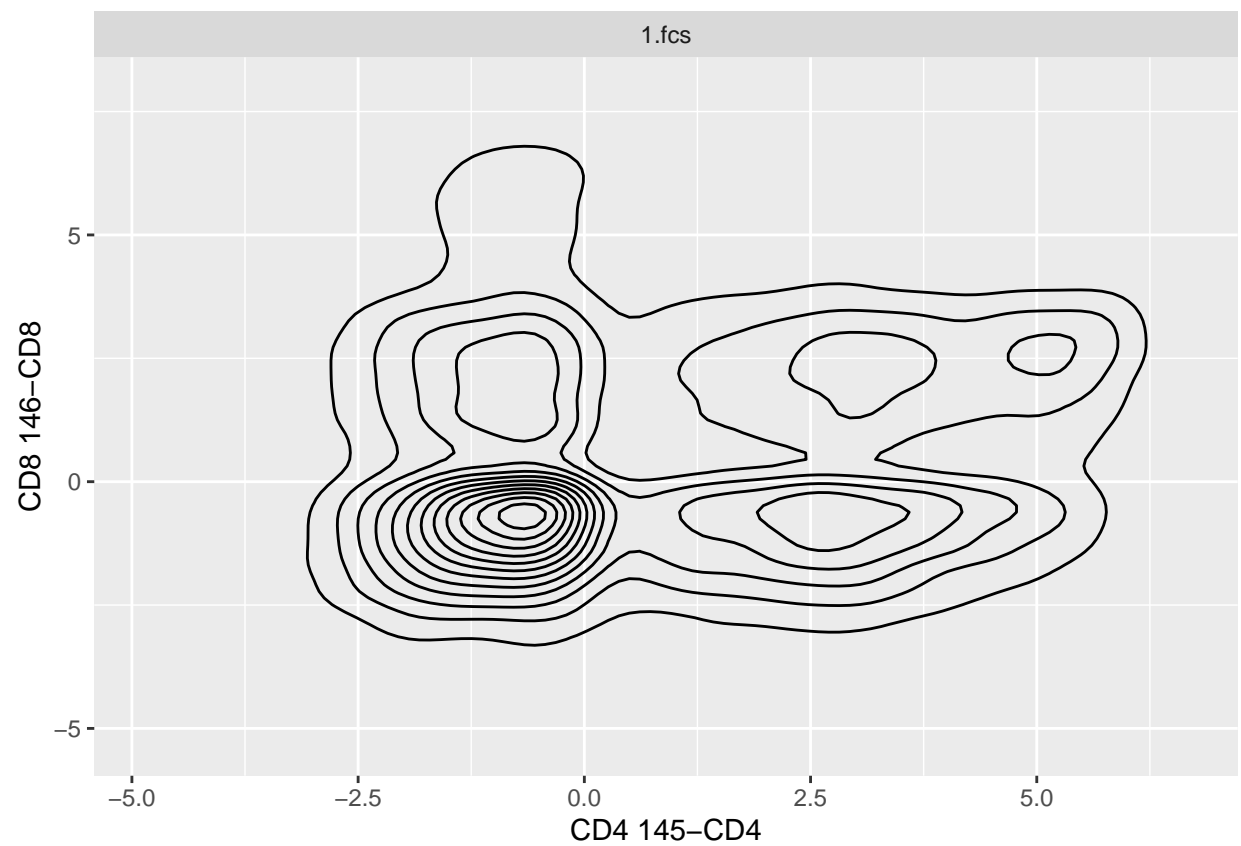
```



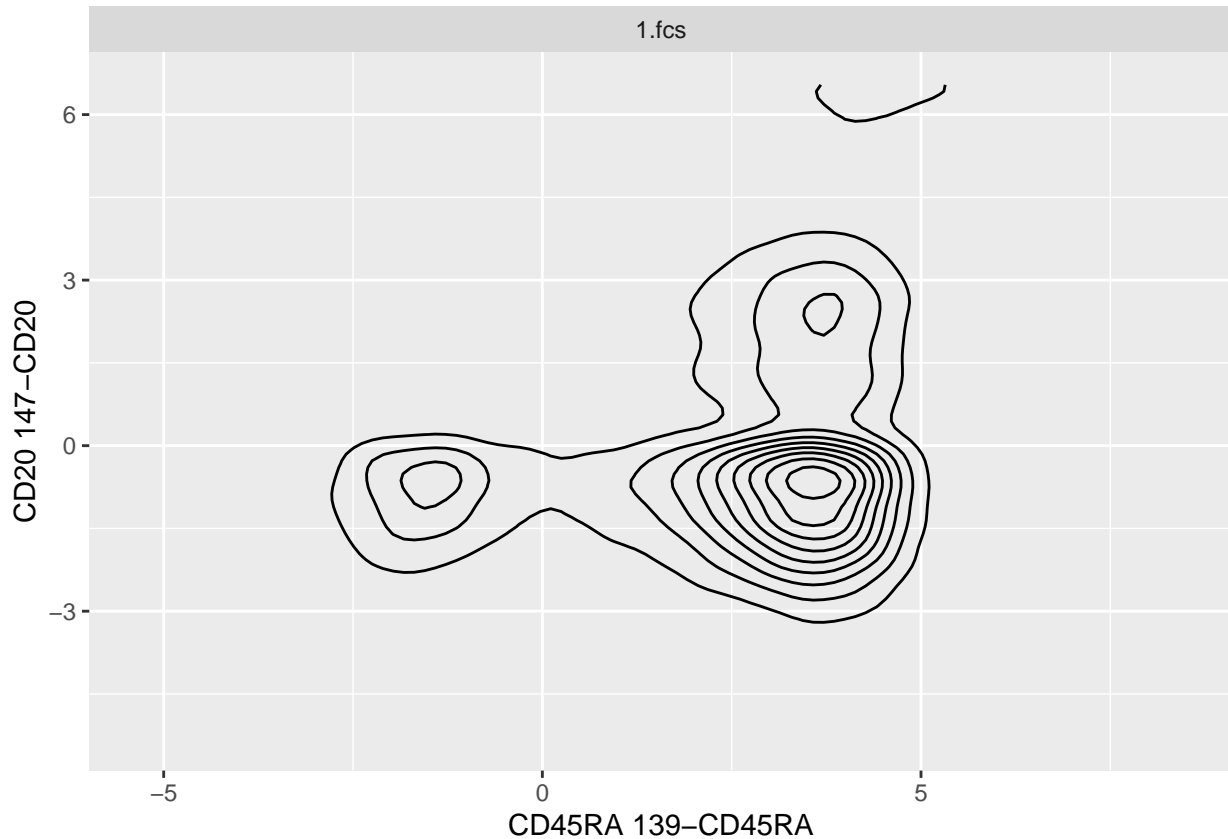
```

p2t <- ggcyto(fcsBT, aes(x=CD4, y = CD8)) +
  geom_density2d(colour="black")
p2t

```



```
p3t <- ggcyto(fcsBT, aes(x = CD45RA, y = CD20)) +  
  geom_density2d(colour = "black")  
p3t
```



5.5.3 Density-based clustering

```
mc5 <- exprs(fcsBT)[,c(15, 16, 19, 40, 33)] #why this order?
res5 <- dbscan::dbscan(mc5, eps = 0.65, minPts = 30)
mc5df <- data.frame(mc5, cluster = as.factor(res5$cluster))
table(mc5df$cluster)
```

```
##
##      0      1      2      3      4      5      6      7      8
## 75954 4031 5450 5310  259  257   63   25   43
```

```
mc6 <- exprs(fcsBT)[,c(15, 16, 19, 40, 25, 33)] #why this order?
res6 <- dbscan::dbscan(mc6, eps = 0.65, minPts = 20)
mc6df <- data.frame(mc6, cluster = as.factor(res6$cluster))
table(mc6df$cluster)
```

```
##
##      0      1      2      3      4      5      6
## 91068   34   61   20   67  121   21
```

```
res6_b <- dbscan::dbscan(mc6, eps = 0.45, minPts = 20)
mc6df_b <- data.frame(mc6, cluster = as.factor(res6_b$cluster))
table(mc6df_b$cluster)
```

```
##
##      0
## 91392
```

```
res6_c <- dbSCAN::dbSCAN(mc6, eps = 0.75, minPts = 20)
mc6df_c <- data.frame(mc6, cluster = as.factor(res6_c$cluster))
table(mc6df_c$cluster)
```

```
##
##      0      1      2      3      4      5      6      7      8      9     10     11
## 88650  412   384   733   170   599    20   167   143    27    28    16
##      12     13
##      23     20
```

Question 5.8

```
load("../..data/Morder.RData")
#without dendrogram or reordering, Euclidean and Manhattan distances
```

```
library(gridExtra)
```

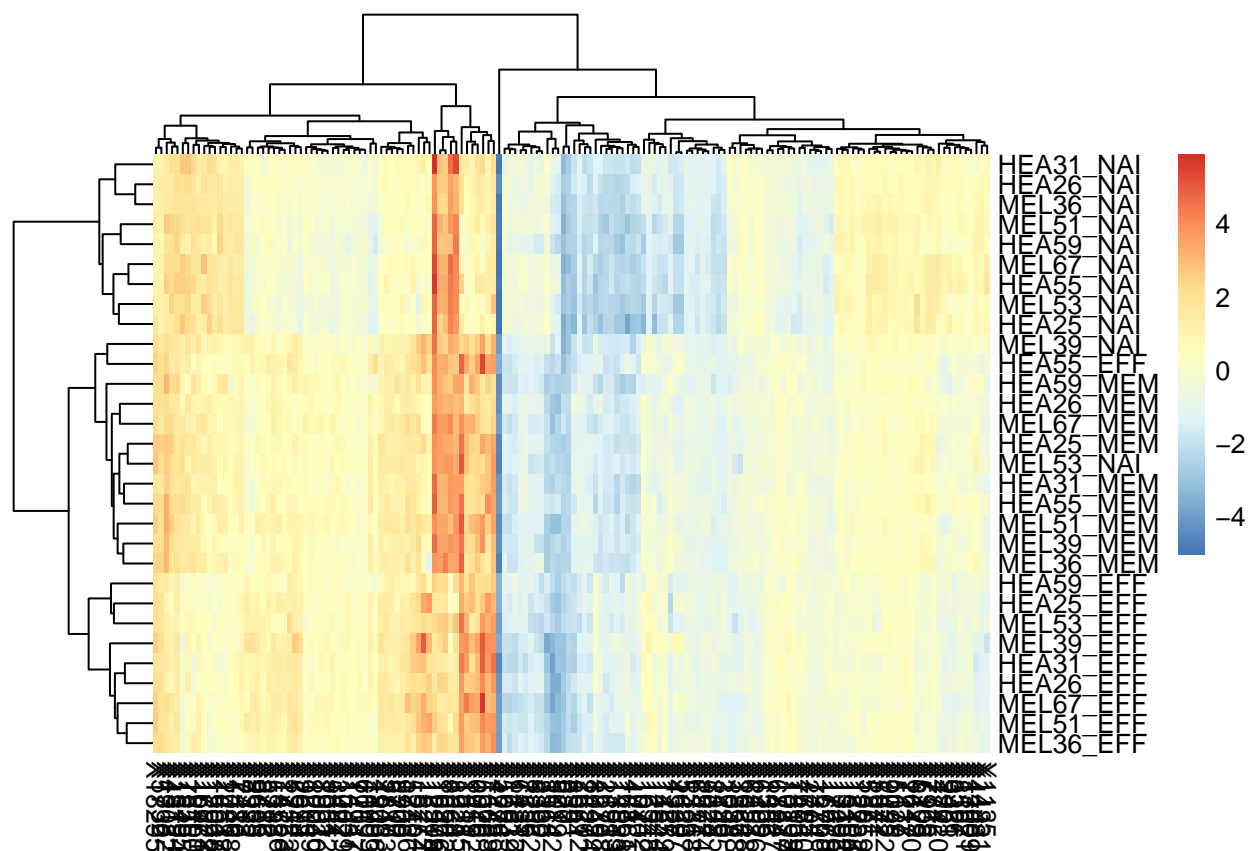
```
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##      combine
```

```
library(grid)
```

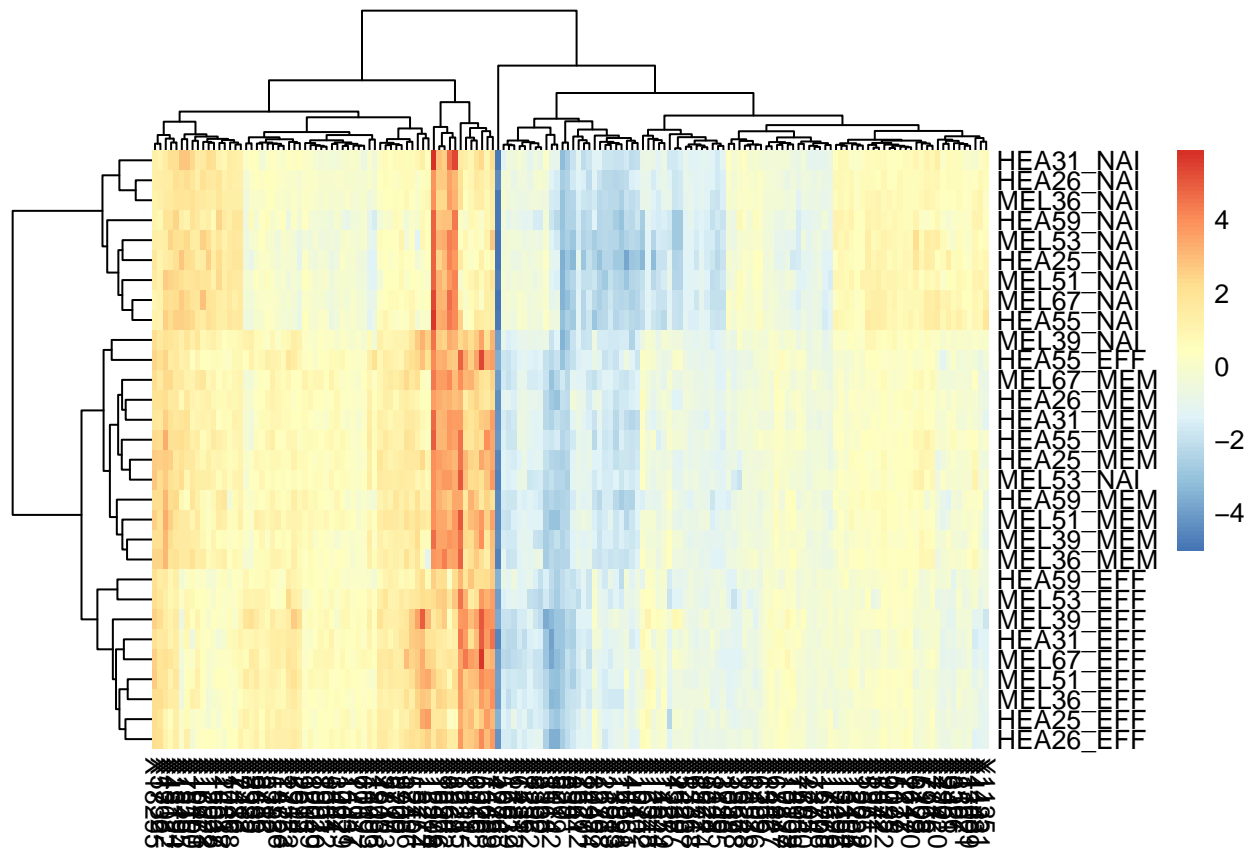
```
##
## Attaching package: 'grid'
## The following object is masked from 'package:mixtools':
##
##      depth
```

```
#I want to plot both on the same panel
#two ways of doing it
#https://www.biostars.org/p/128229/
#https://stackoverflow.com/questions/39590849/using-a-heatmap-in-arrangegrob
```

```
plot_list <- list()
ph1 <- pheatmap(Morder)
```

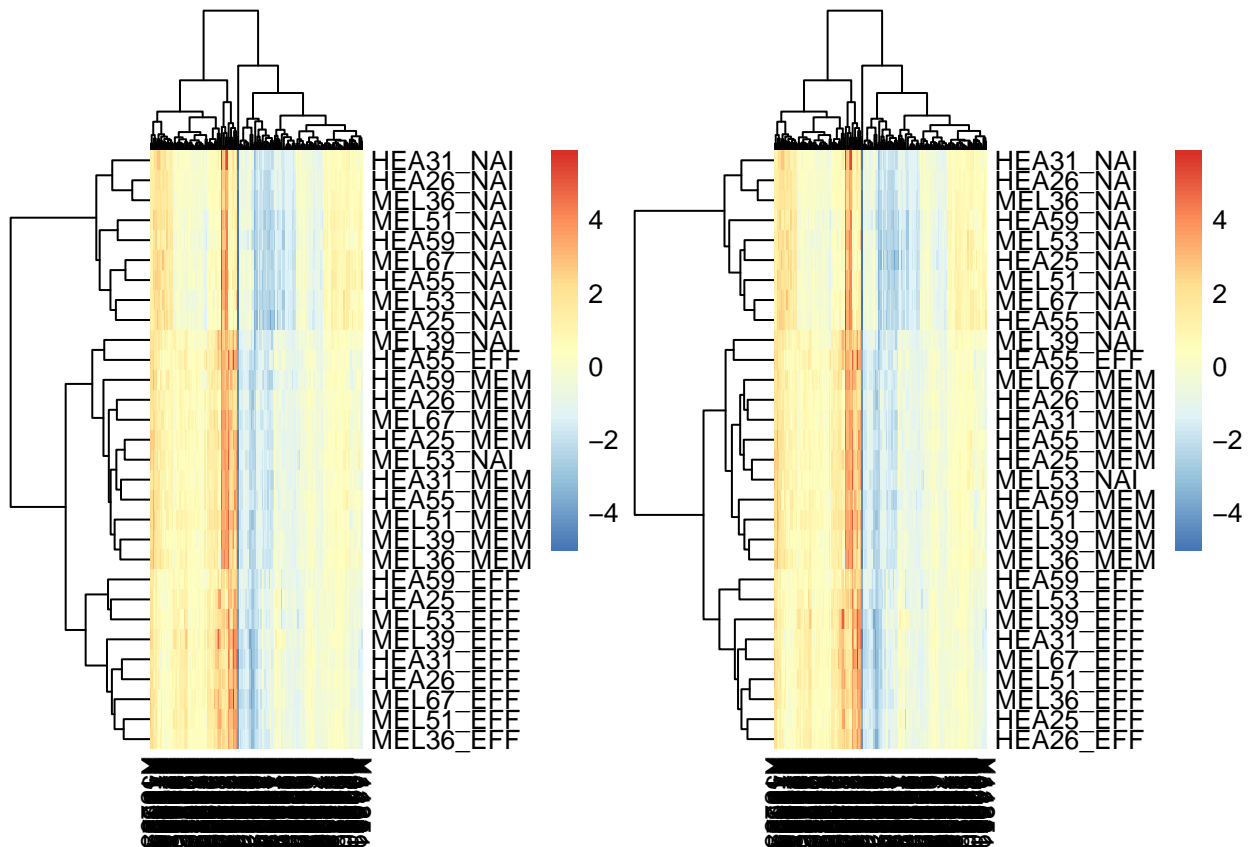


```
ph2 <- pheatmap(Morder, clustering_distance_rows = "manhattan")
```



```
plot_list[[1]] <- ph1[[4]]
plot_list[[2]] <- ph2[[4]]

g <- grid.arrange(arrangeGrob(grobs= plot_list,ncol=2))
```

```
#g<-do.call(grid.arrange,plot_list)
```

```
#Question 5.9: which orderings do not match
```

```
#https://www.biostars.org/p/170614/
```

```
res1 <- Morder[c(ph1$tree_row[["order"]]),ph1$tree_col[["order"]]]
```

```
res2 <- Morder[c(ph2$tree_row[["order"]]),ph2$tree_col[["order"]]]
```

```
row.names(res1) == row.names(res2)
```

```
## [1] TRUE TRUE TRUE FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE
## [12] FALSE TRUE FALSE FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE
## [23] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
```

here

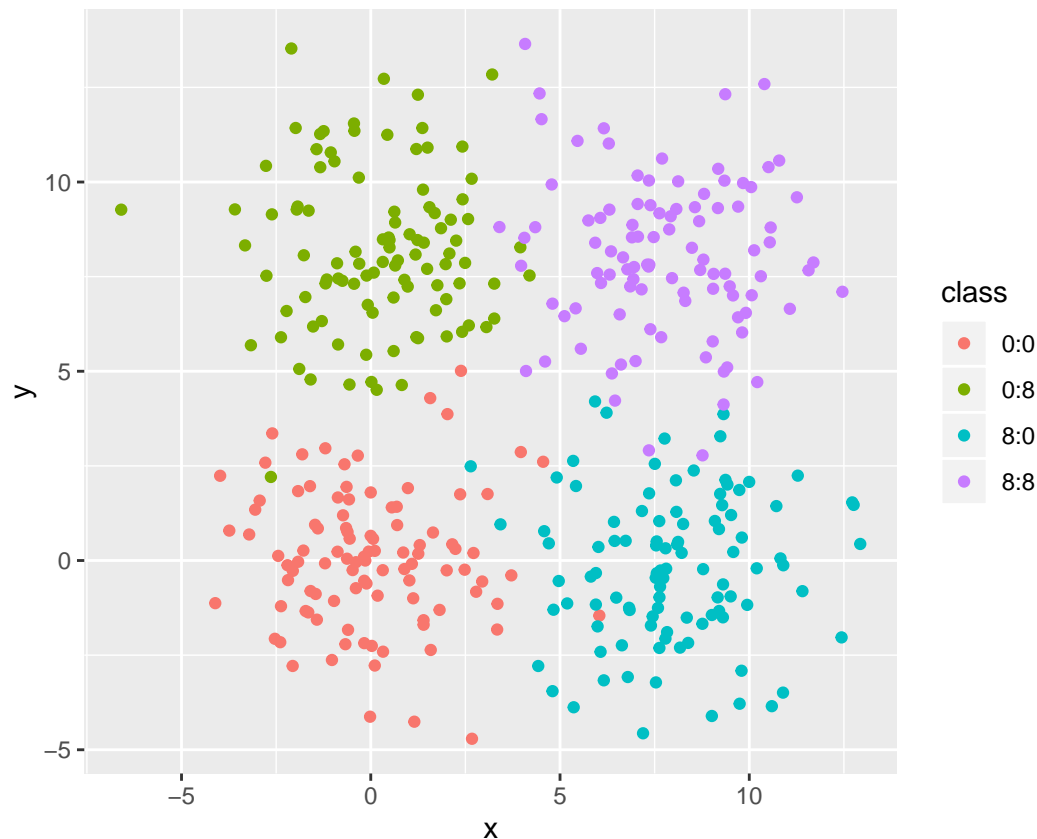
Question 5.12

```
library(tidyverse)
simdat <- lapply(c(0, 8), function(mx){
  lapply(c(0, 8), function(my){
    tibble(x = rnorm(100, mean = mx, sd = 2),
           y = rnorm(100, mean = my, sd = 2),
           class = paste(mx, my, sep=":"))
  }) %>% bind_rows
}) %>% bind_rows
simdat
```

```
## # A tibble: 400 x 3
```

```
##      x      y class
##    <dbl> <dbl> <chr>
## 1 -1.20  2.97  0:0
## 2 -2.53 -2.07  0:0
## 3 -2.79  2.58  0:0
## 4 -3.74  0.791 0:0
## 5  3.71 -0.392 0:0
## 6 -0.575 1.61  0:0
## 7  1.26  0.180 0:0
## 8  2.94 -0.554 0:0
## 9 -1.92 -0.0333 0:0
## 10 -0.172 -0.535 0:0
## # ... with 390 more rows
```

```
simdatxy <- simdat[, c("x", "y")]
ggplot(simdat, aes(x = x, y = y, col=class))+
  geom_point()+
  coord_fixed()
```



```
rsm1 <- range(c(simdat$x[1:100], simdat$y[1:100]))
rsm2 <- range(c(simdat$x[101:200], simdat$y[101:200]))
rsm3 <- range(c(simdat$x[201:300], simdat$y[201:300]))
rsm4 <- range(c(simdat$x[301:400], simdat$y[301:400]))

simdat_un <- lapply(c(0, 8), function(mx){
  lapply(c(0, 8), function(my){
    tibble(x = runif(100, range(simdat$x)[1], range(simdat$x)[2]),
           y = runif(100, range(simdat$y)[1], range(simdat$y)[2]),
```

```

      class = paste(mx, my, sep=":")
    }) %>% bind_rows
  }) %>% bind_rows
  simdat_un

```

```

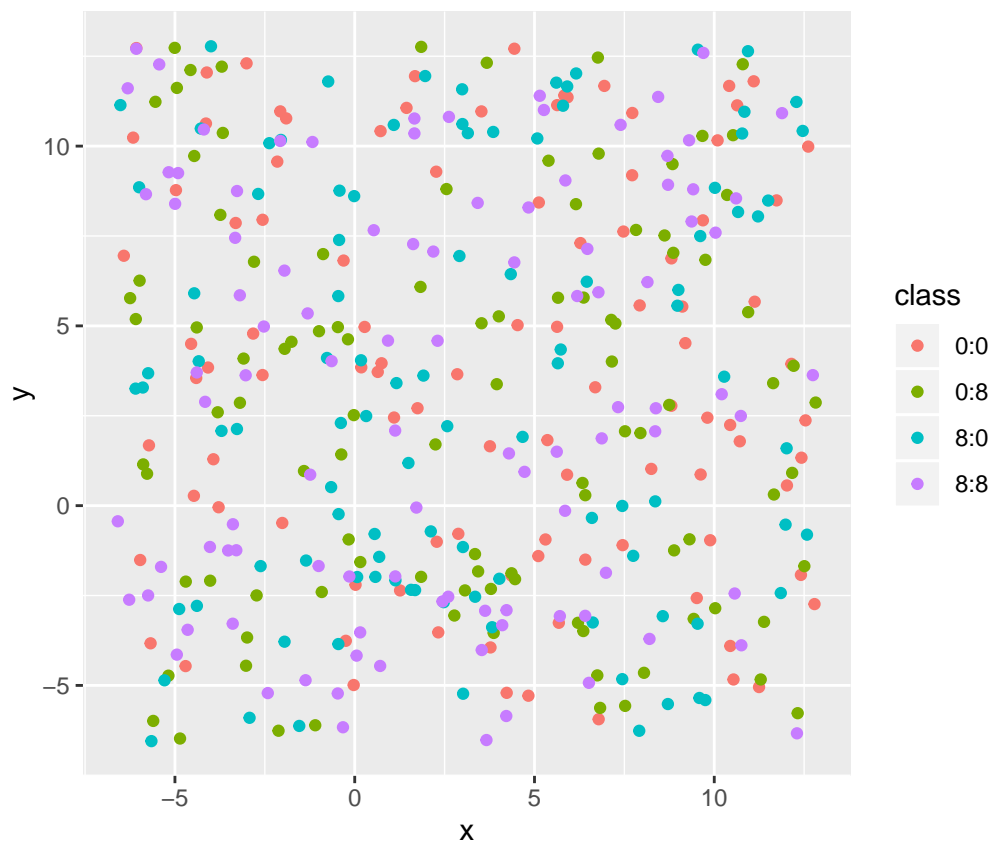
## # A tibble: 400 x 3
##       x       y class
##   <dbl> <dbl> <chr>
## 1  5.31 -0.941 0:0
## 2  8.25  1.02  0:0
## 3 12.6   9.99  0:0
## 4 11.7   8.49  0:0
## 5  0.276 4.97  0:0
## 6 -4.55  4.50  0:0
## 7  5.12  8.43  0:0
## 8 11.1  11.8  0:0
## 9 -1.91 10.8  0:0
## 10 7.45 -1.10 0:0
## # ... with 390 more rows

```

```

ggplot(simdat_un, aes(x = x, y = y, col=class))+
  geom_point()+
  coord_fixed()

```



```

library(cluster)

pamfun <- function(x, k)list(cluster=pam(x, k, cluster.only = TRUE))

```

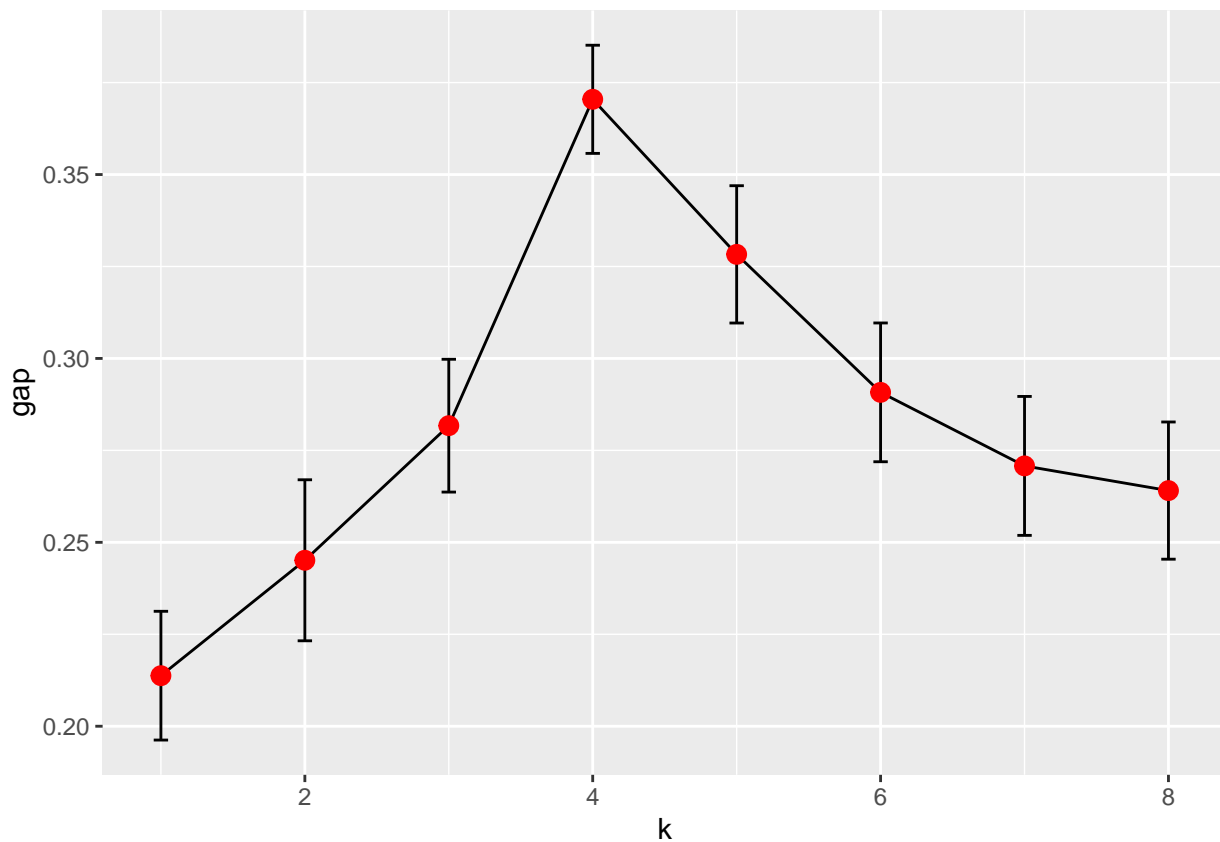
```

gss <- clusGap(simdatxy, FUN = pamfun, K.max = 8, B = 50, verbose = FALSE)

plot_gap <- function(x){
  gstab <- data.frame(x$Tab, k = seq_len(nrow(x$Tab)))
  ggplot(gstab, aes(k, gap)) +
    geom_line() +
    geom_errorbar(aes(ymax = gap + SE.sim,
                     ymin = gap - SE.sim), width = 0.1) +
    geom_point(size = 3, col = "red")
}

plot_gap(gss)

```



```

library("Hiiragi2013")

## Loading required package: affy
## Loading required package: BiocGenerics
## Loading required package: parallel
##
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:parallel':
##
##   clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
##   clusterExport, clusterMap, parApply, parCapply, parLapply,
##   parLapplyLB, parRapply, parSapply, parSapplyLB

```

```

## The following object is masked from 'package:gridExtra':
##
##   combine
## The following object is masked from 'package:flowCore':
##
##   normalize
## The following objects are masked from 'package:dplyr':
##
##   combine, intersect, setdiff, union
## The following objects are masked from 'package:stats':
##
##   IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
##   anyDuplicated, append, as.data.frame, basename, cbind,
##   colnames, dirname, do.call, duplicated, eval, evalq, Filter,
##   Find, get, grep, grepl, intersect, is.unsorted, lapply, Map,
##   mapply, match, mget, order, paste, pmax, pmax.int, pmin,
##   pmin.int, Position, rank, rbind, Reduce, rownames, sapply,
##   setdiff, sort, table, tapply, union, unique, unsplit, which,
##   which.max, which.min
## Loading required package: Biobase
## Welcome to Bioconductor
##
##   Vignettes contain introductory material; view with
##   'browseVignettes()'. To cite Bioconductor, see
##   'citation("Biobase")', and for packages 'citation("pkgname)".
## Loading required package: boot
##
## Attaching package: 'boot'
## The following object is masked from 'package:lattice':
##
##   melanoma
## Loading required package: clue
## Loading required package: genefilter
##
## Attaching package: 'genefilter'
## The following object is masked from 'package:readr':
##
##   spec
## Loading required package: geneplotter
## Loading required package: annotate
## Loading required package: AnnotationDbi
## Loading required package: stats4
## Loading required package: IRanges

```

```

## Loading required package: S4Vectors
##
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:tidyr':
##
##     expand
## The following objects are masked from 'package:dplyr':
##
##     first, rename
## The following object is masked from 'package:base':
##
##     expand.grid
##
## Attaching package: 'IRanges'
## The following object is masked from 'package:purrr':
##
##     reduce
## The following objects are masked from 'package:dplyr':
##
##     collapse, desc, slice
##
## Attaching package: 'AnnotationDbi'
## The following object is masked from 'package:dplyr':
##
##     select
## Loading required package: XML
##
## Attaching package: 'annotate'
## The following object is masked from 'package:flowCore':
##
##     journal
## Loading required package: gplots
##
## Attaching package: 'gplots'
## The following object is masked from 'package:IRanges':
##
##     space
## The following object is masked from 'package:S4Vectors':
##
##     space
## The following object is masked from 'package:stats':
##
##     lowess
## Loading required package: gtools

```

```

##
## Attaching package: 'gtools'

## The following objects are masked from 'package:boot':
##
##     inv.logit, logit

## The following object is masked from 'package:mixtools':
##
##     ddirichlet

## Loading required package: KEGGREST

## Loading required package: MASS

##
## Attaching package: 'MASS'

## The following object is masked from 'package:AnnotationDbi':
##
##     select

## The following object is masked from 'package:genefilter':
##
##     area

## The following object is masked from 'package:dplyr':
##
##     select

## Loading required package: mouse4302.db

## Loading required package: org.Mm.eg.db

##
##

## Loading required package: RColorBrewer

## Loading required package: xtable

data("x")

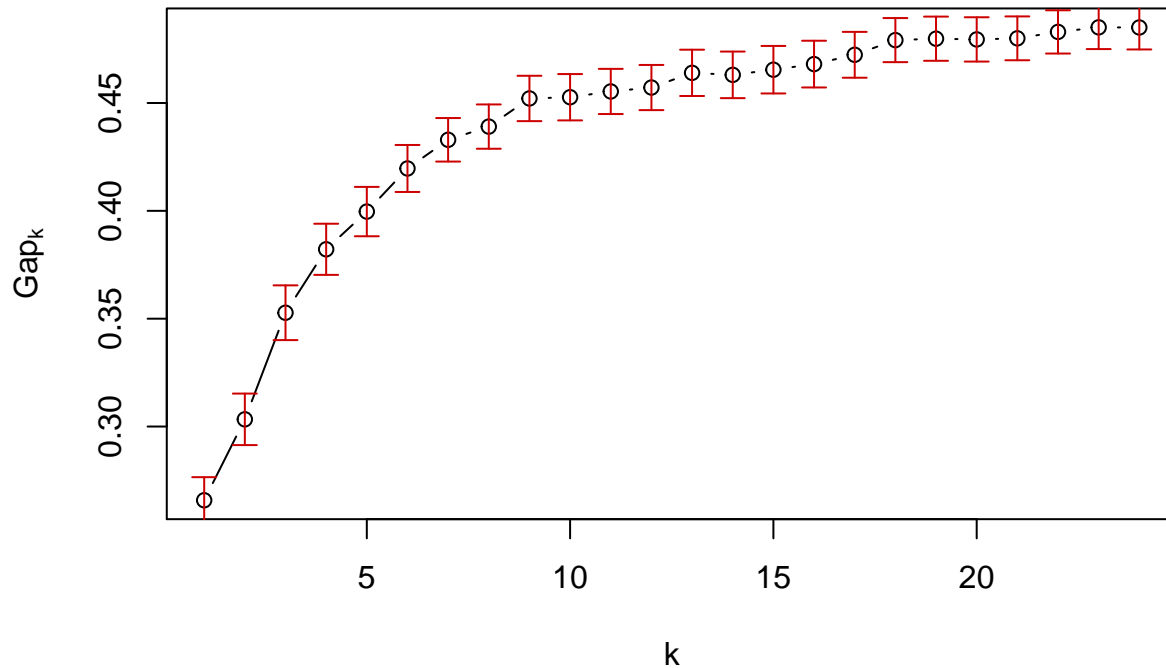
selFeats = order(rowVars(Biobase::exprs(x)), decreasing = TRUE)[1:50]
embmat = t(Biobase::exprs(x)[selFeats, ])
embgap = clusGap(embmat, FUN = pamfun, K.max = 24, verbose = FALSE)
k1 = maxSE(embgap$Tab[, "gap"], embgap$Tab[, "SE.sim"])
k2 = maxSE(embgap$Tab[, "gap"], embgap$Tab[, "SE.sim"],
           method = "Tibs2001SEmax")
c(k1, k2)

## [1] 11 7

plot(embgap)

```

**clusGap(x = embmat, FUNcluster = pamfun, K.max = 24,
verbose = FALSE)**



Question 5.15 Use all features in x #times out

```
embmat_full = t(Biobase::exprs(x))
embgap_full = clusGap(embmat_full, FUN = pamfun, K.max = 24, verbose = FALSE)
k1 = maxSE(embgap_full$Tab[, "gap"], embgap_full$Tab[, "SE.sim"])
k2 = maxSE(embgap_full$Tab[, "gap"], embgap_full$Tab[, "SE.sim"],
           method = "Tibs2001SEmax")
c(k1, k2)
```

```
plot(embgap)
```

```
clusterResampling = function(x, ngenes = 50, k = 2, B = 250,
                             prob = 0.67) {
```

```
  mat = Biobase::exprs(x)
```

```
  #draw a random resampling of 67% of the data without replacement
```

```
  #repeat B times
```

```
  ce = cl_ensemble(list = lapply(seq_len(B), function(b) {
```

```
    selSamps = sample(ncol(mat), size = round(prob * ncol(mat)),
                     replace = FALSE)
```

```
    submat = mat[, selSamps, drop = FALSE]
```

```
    #select the top n genes features by overall variance in the subset
```

```
    sel = order(rowVars(submat), decreasing = TRUE)[seq_len(ngenes)]
```

```
    submat = submat[sel, , drop = FALSE]
```

```
    #apply kmeans clustering
```

```
    pamres = pam(t(submat), k = k)
```

```
    #predict cluster memberships of the samples that were not in the subset with the cl_predict method
```

```
    pred = cl_predict(pamres, t(mat[sel, ]), "memberships")
```

```
    as.cl_partition(pred)
```

```
  }))
```

```
  #for each of B clusterings, measure the agreement with the consensus through the function cl_agreement
```



```

cons = cl_consensus(ce)
ag = sapply(ce, cl_agreement, y = cons)
list(agreements = ag, consensus = cons)
}

iswt <- (x$genotype == "WT")
cr1 <- clusterResampling(x[, x$Embryonic.day == "E3.25" & iswt])
cr2 <- clusterResampling(x[, x$Embryonic.day == "E3.5" & iswt])

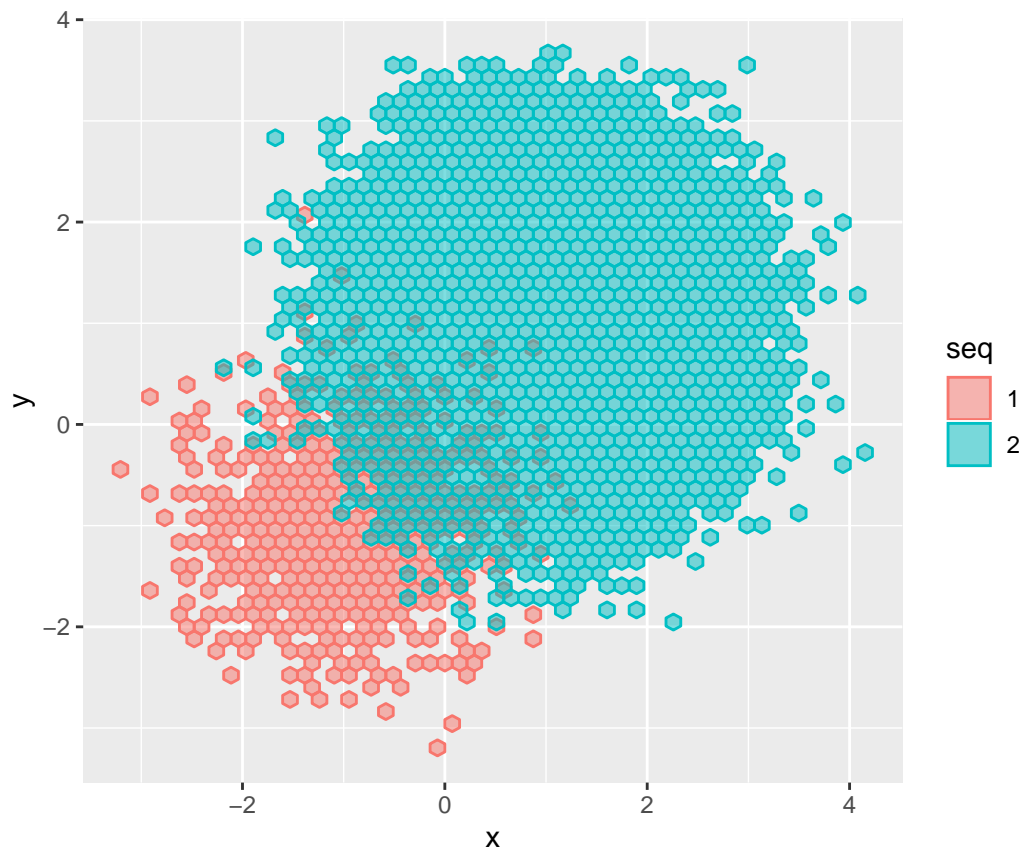
```

5.8 Clustering as a means for denoising

```

library("mixtools")
library("ggplot2")
seq1 = rmvnorm(n = 1e3, mu = -c(1, 1), sigma = 0.5 * diag(c(1, 1)))
seq2 = rmvnorm(n = 1e5, mu = c(1, 1), sigma = 0.5 * diag(c(1, 1)))
twogr = data.frame(
  rbind(seq1, seq2),
  seq = factor(c(rep(1, nrow(seq1)),
                 rep(2, nrow(seq2))))
)
colnames(twogr)[1:2] = c("x", "y")
ggplot(twogr, aes(x = x, y = y, colour = seq, fill = seq)) +
  geom_hex(alpha = 0.5, bins = 50) + coord_fixed()

```



Question 5.16: Take the data seq1 and seq2 and cluster them into two groups according to distance from the

group center

Exercises

Exercise 5.1

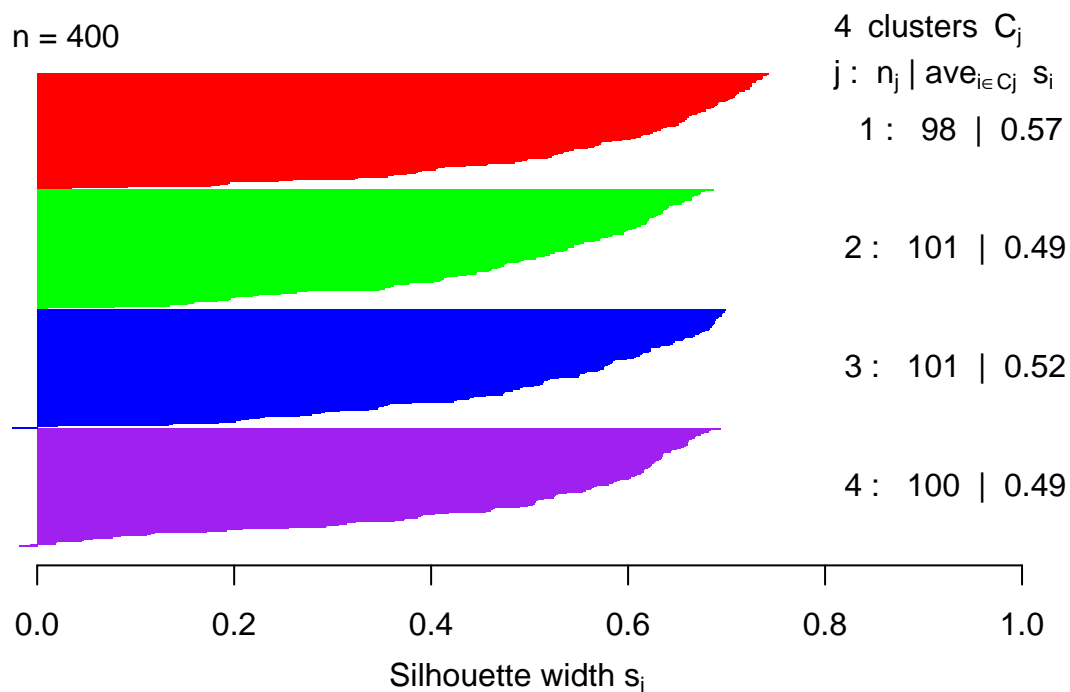
```
library("cluster")
library("viridis")

## Loading required package: viridisLite

pam4 <- pam(simdatxy, 4)
sil <- silhouette(pam4, 8)
plot(sil, col=c("red", "green", "blue", "purple"), main = "Silhouette")
```

Silhouette

n = 400



Average silhouette width : 0.52

```
# coloursV <- viridis(10, direction=-1)
# for(k in 2:10)
#   plot(silhouette(pam(simdatxy, k=k)), main = paste("k = ",k), do.n.k=FALSE,
#        col = coloursV[1:k])

sil_index <- c()
for(k in 2:20){
  pam4 <- pam(simdatxy, k)
  sil <- silhouette(pam4, k)
  sil_index[k-1] <- mean(sil[,3])
}
```

```
sil_index_un <- c()
for(k in 2:20){
  pam4 <- pam(simdat_un, k)
  sil <- silhouette(pam4, k)
  sil_index_un[k-1] <- mean(sil[,3])
}
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

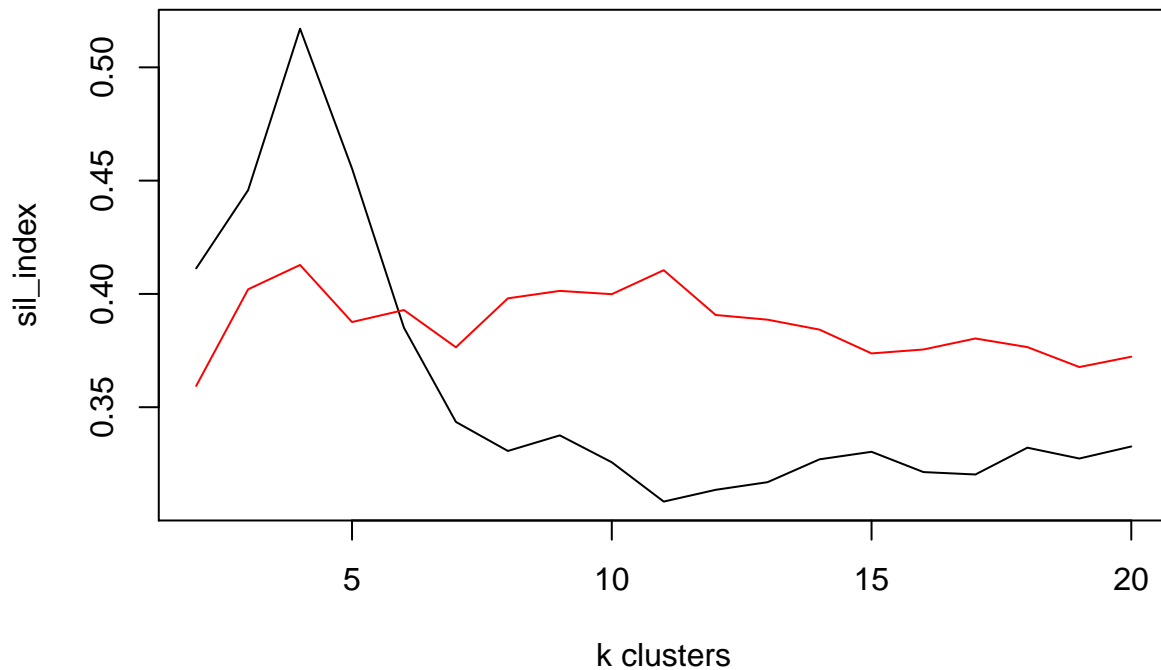
```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
## Warning in data.matrix(x): NAs introduced by coercion
```

```
plot(2:20, sil_index, type="l", xlab = "k clusters")
```

```
points(2:20, sil_index_un, type="l", col="red")
```



Exercise 5.2

```
library(vegan)
```

```
## Loading required package: permute
```

```
##
```

```
## Attaching package: 'permute'
```

```
## The following object is masked from 'package:gtools':
```

```
##
```

```
## permute
```

```
## The following object is masked from 'package:devtools':
```

```
##
```

```
## check
```

```
## This is vegan 2.5-6
```

```
data(dune)
```

```
dune_cor = cor(dune)
```

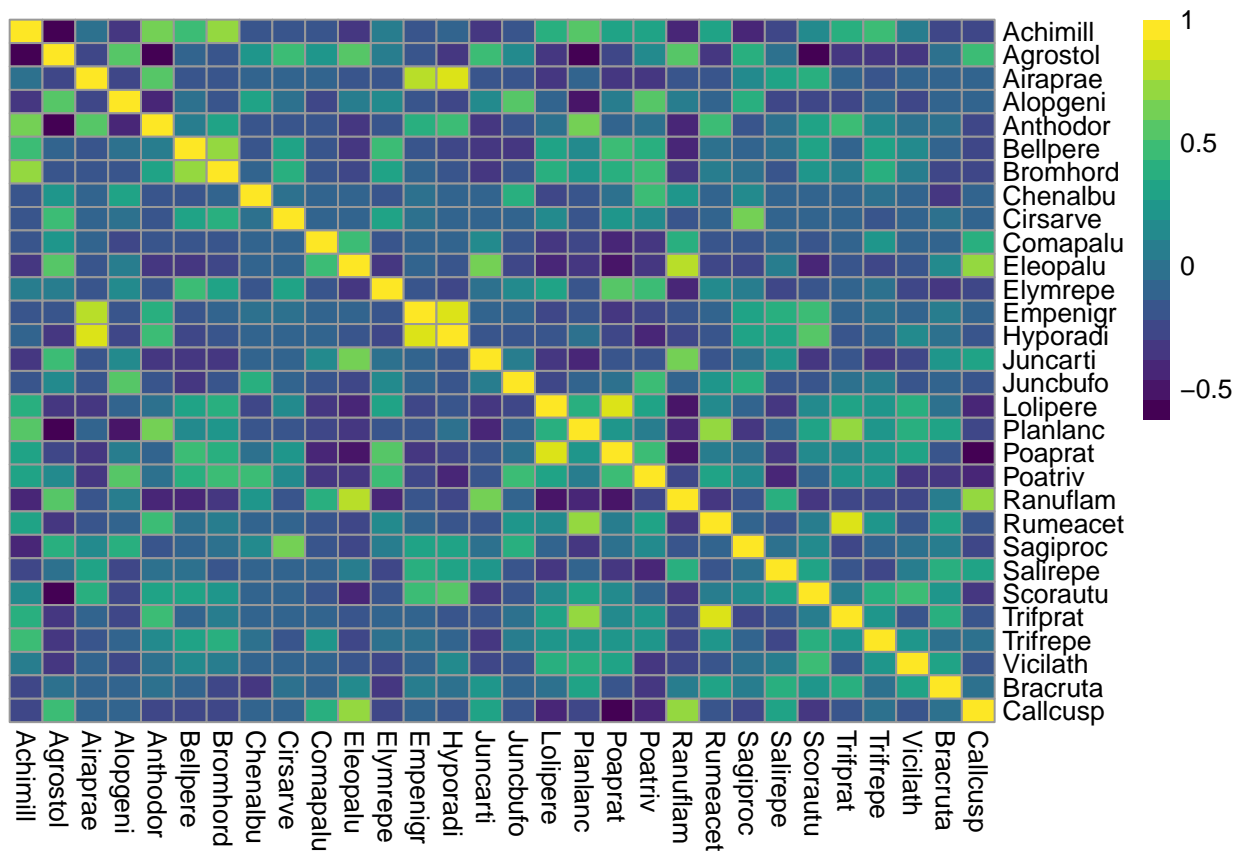
```
symnum(dune_cor)
```

```
##          Ac Ag Ar Al An Bl Brm Ch Cr Cm Elp Ely Em H Jncr Jncb L Pl Pp Pt
## Achimill 1
## Agrostol , 1
## Airaprae      1
## Alop geni . .      1
## Anthodor , . . . 1
## Bellpere .          1
## Bromhord ,          , 1
## Chenalbu          , 1
## Cirsarve .          . .      1
## Comapalu          , 1
## Eleopal . .          . .      1
## Elymrepe .          . .      1
```

```

## Empenigr      +      .      1
## Hyporadi      . *      .      + 1
## Juncarti . . . . . ,      1
## Juncbufo      . . . . .      1
## Lolipere . . . . . . . . 1
## Planlanc . , . . . . . . 1
## Poaprat . . . . . . . . + 1
## Poatriv      . . . . . . . . 1
## Ranuflam . . . . . , . . .
## Rumeacet . . . . . , . .
## Sagiproc . . . . . , . . .
## Salirepe      . . . . . . . .
## Scorausu . . . . . . . .
## Trifprat . . . . . . . .
## Trifrepe . . . . . . . .
## Vicilath . . . . . . . .
## Bracruta      . . . . . . . .
## Callcusp . . . . . , . . .
##      Rn Rm Sg Sl Sc Trfp Trfr V Brc Cl
## Achimill
## Agrostol
## Airaprae
## Alop geni
## Anthodor
## Bellpere
## Bromhord
## Chenalbu
## Cirsarve
## Comapalu
## Eleopalu
## Elymrepe
## Empenigr
## Hyporadi
## Juncarti
## Juncbufo
## Lolipere
## Planlanc
## Poaprat
## Poatriv
## Ranuflam 1
## Rumeacet . 1
## Sagiproc      1
## Salirepe .      1
## Scorausu . . 1
## Trifprat +      1
## Trifrepe      .      1
## Vicilath      .      1
## Bracruta . . . . . 1
## Callcusp , . . . . 1
## attr("legend")
## [1] 0 ' ' 0.3 '.' 0.6 ',' 0.8 '+' 0.9 '*' 0.95 'B' 1
pheatmap(dune_cor, cluster_rows = FALSE, cluster_cols = FALSE, color = viridis(20))

```



Exercise 5.3

```
library(kernlab)
```

```
##
## Attaching package: 'kernlab'
## The following object is masked from 'package:permute':
##
##   how
## The following object is masked from 'package:purrr':
##
##   cross
## The following object is masked from 'package:ggplot2':
##
##   alpha
```

```
data(spirals)
```