

TFM_E2_Grupo02

2023-06-08

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':  
##  
##   filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
##   intersect, setdiff, setequal, union
```

```
library(factoextra)
```

```
## Loading required package: ggplot2
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(ggplot2)  
library(cluster)  
library(tidyverse)
```

```
## — Attaching core tidyverse packages ————— tidyverse 2.0.0 —  
## ✓ forcats   1.0.0   ✓ stringr  1.5.0  
## ✓ lubridate 1.9.2   ✓ tibble  3.2.1  
## ✓ purrr     1.0.1   ✓ tidyr   1.3.0  
## ✓ readr     2.1.4
```

```
## — Conflicts ————— tidyverse_conflicts() —  
## ✗ dplyr::filter() masks stats::filter()  
## ✗ dplyr::lag()    masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(ggcorrplot)  
library(writexl)
```

```
## Warning: package 'writexl' was built under R version 4.3.1
```

```
library(openxlsx)
```

```
## Warning: package 'openxlsx' was built under R version 4.3.1
```

```
library(PerformanceAnalytics)
```

```
## Warning: package 'PerformanceAnalytics' was built under R version 4.3.1
```

```
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric
##
##
## ##### Warning from 'xts' package #####
## #
## # The dplyr lag() function breaks how base R's lag() function is supposed to #
## # work, which breaks lag(my_xts). Calls to lag(my_xts) that you type or #
## # source() into this session won't work correctly. #
## # #
## # Use stats::lag() to make sure you're not using dplyr::lag(), or you can add #
## # conflictRules('dplyr', exclude = 'lag') to your .Rprofile to stop #
## # dplyr from breaking base R's lag() function. #
## # #
## # Code in packages is not affected. It's protected by R's namespace mechanism #
## # Set `options(xts.warn_dplyr_breaks_lag = FALSE)` to suppress this warning. #
## # #
## #####
##
## Attaching package: 'xts'
##
## The following objects are masked from 'package:dplyr':
##
##   first, last
##
##
## Attaching package: 'PerformanceAnalytics'
##
## The following object is masked from 'package:graphics':
##
##   legend
```

```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
#cargamos los datos con las variables que nos interesan
#df1 = data.frame(read.csv2('C:/Users/Linfante/OneDrive/Documentos/MasterBigData/TFM/tfm_master_
ds/1 - BackEnd/Datos/alimentos v3.csv', sep=','))
#df2 = data.frame(read.csv2('C:/Users/Linfante/OneDrive/Documentos/MasterBigData/TFM/tfm_master_
ds/1 - BackEnd/Datos/alimentos_vegan.csv', sep=','))
df = data.frame(read.csv2('C:/Users/Linfante/OneDrive/Documentos/MasterBigData/TFM/tfm_master_d
s/1 - BackEnd/Datos/alimentos v4.csv', sep=','))
```

```
#df <- df[,c("PRODUCT_NAME", "PROTEINS_100G", "CARBOHYDRATES_100G", "FAT_100G", "FIBER_100G", "SALT_1
00G", "SATURATED_FAT_100G", "SUGARS_100G")]
df <- df[,c("PRODUCT_NAME", "PROTEINS_100G", "CARBOHYDRATES_100G", "FAT_100G")]
```

```
PRODUCT_NAME <- df[,1]
```

```
# Filtrar las columnas que contienen el texto "_100G"
```

```
#columnas_filtradas <- grep("_100G", colnames(df))
```

```
#df <- df[, columnas_filtradas]
```

```
#df <- cbind(PRODUCT_NAME, df)
```

```
#Convertir a numerico todo
```

```
df_nuevo <- df %>%
```

```
  mutate(across(where(is.character), type.convert, as.is = TRUE)) %>%
```

```
  select_if(is.numeric)
```

```
## Warning: There was 1 warning in `mutate()`.
```

```
## i In argument: `across(where(is.character), type.convert, as.is = TRUE)`.
```

```
## Caused by warning:
```

```
## ! The `...` argument of `across()` is deprecated as of dplyr 1.1.0.
```

```
## Supply arguments directly to `.fns` through an anonymous function instead.
```

```
##
```

```
## # Previously
```

```
## across(a:b, mean, na.rm = TRUE)
```

```
##
```

```
## # Now
```

```
## across(a:b, \(x) mean(x, na.rm = TRUE))
```

```
total <- as.data.frame(colSums(df_nuevo))
```

```
totalc <- as.data.frame(rowSums(df_nuevo))
```

```

totales <- colSums(df_nuevo)

# Seleccionar las columnas con un total mayor o igual a 5
#columnas_a_mantener <- totales >= 10000

# Eliminar las columnas con un total inferior a 30
#df_nuevo <- df_nuevo[, columnas_a_mantener]
PRODUCT_NAME <- df[,1]
df_nuevo <- cbind(PRODUCT_NAME,df_nuevo)

```

```

#elimina vectores fila nulos
df_nuevo$suma <- rowSums(df_nuevo[, c("PROTEINS_100G", "CARBOHYDRATES_100G", "FAT_100G")])
df_nuevo <- subset(df_nuevo, !(suma == 0))
df_nuevo <- df_nuevo[,c("PRODUCT_NAME", "PROTEINS_100G", "CARBOHYDRATES_100G", "FAT_100G")]
df_nuevo <- na.omit(df_nuevo)

```

```

#filtros negativos
# Crear un vector con las palabras a buscar
palabras_a_eliminar <- c("SALSA DE SOJA", "SAUCE SOJA", "SAUCE SOJA", "BEBIDA DE SOJA", "YAOURT SOJA", "YOGURT DE SOJA", "CHOCOLAT", "ARROZ Y SOJA", "LECHE", "MOUSSE", "MILK", "ACEITE", "DESSERT", "PAN S OJA", "BOISSON SOJA", "GLACE", "ML", "SAUCE DE SOJA", "SAUCE", "SALSA", "BIBEDA DE SOJA", "LAIT SOJ A", "VIVESOY SOJA", "LAIT DE SOJA", "YAOURT", "POSTRE", "MUESLI", "YOGUR", "BEBIDA", "MARGARINA", "VINAIG RETTE", "DRINK", "SAUCE", "BATIDO", "LAIT", "MAYONNAISE", "CAFE", "VANILLE", "NATA", "YOGURT")

# Crear una función que verifica si alguna de las palabras está presente en el texto
verificar_palabras <- function(texto, palabras) {
  sapply(palabras, function(palabra) grepl(palabra, texto))
}

# Identificar las filas que contienen alguna de las palabras a eliminar
filas_a_eliminar <- apply(df_nuevo, 1, function(row) any(verificar_palabras(row["PRODUCT_NAME"], palabras_a_eliminar)))

# Eliminar las filas identificadas del dataframe original
df_nuevo1 <- df_nuevo[!filas_a_eliminar, ]
#df_nuevo2 <- df_nuevo1[,-1]

#Elimino los faltantes
df_nuevo1 <- df_nuevo1[complete.cases(df_nuevo1), ]

```

```

df_soja <- subset(df_nuevo1, grepl("soja", PRODUCT_NAME, ignore.case = TRUE)) ## Registros que c
ontienen la palabra "soja"
df_seitan <- subset(df_nuevo1, grepl("seitan", PRODUCT_NAME, ignore.case = TRUE)) ## Registros q
ue contienen la palabra "seitan"
df_tofu <- subset(df_nuevo1, grepl("tofu", PRODUCT_NAME, ignore.case = TRUE)) ## Registros que c
ontienen la palabra "tofu"
df_tofu <- na.omit(df_tofu)

```

```
# Definir la función para eliminar valores atípicos basados en desviación estándar
remove_outliers <- function(data, column, sd_threshold = 2) {
  data[abs(scale(data[[column]])) < sd_threshold, ]
}

# Eliminar valores atípicos en columna1 utilizando desviación estándar
df_soja <- remove_outliers(df_soja, "PROTEINS_100G")
df_soja <- remove_outliers(df_soja, "CARBOHYDRATES_100G")
df_soja <- remove_outliers(df_soja, "FAT_100G")
#df_soja <- remove_outliers(df_soja, "SUGARS_100G")

# Eliminar valores atípicos en columna1 utilizando desviación estándar
df_seitan <- remove_outliers(df_seitan, "PROTEINS_100G")
df_seitan <- remove_outliers(df_seitan, "CARBOHYDRATES_100G")
df_seitan <- remove_outliers(df_seitan, "FAT_100G")
#df_seitan <- remove_outliers(df_seitan, "SUGARS_100G")

# Eliminar valores atípicos en columna1 utilizando desviación estándar
df_tofu <- remove_outliers(df_tofu, "PROTEINS_100G")
df_tofu <- remove_outliers(df_tofu, "CARBOHYDRATES_100G")
df_tofu <- remove_outliers(df_tofu, "FAT_100G")
#df_tofu <- remove_outliers(df_tofu, "SUGARS_100G")
```

Soja

```
summary(df_soja)
```

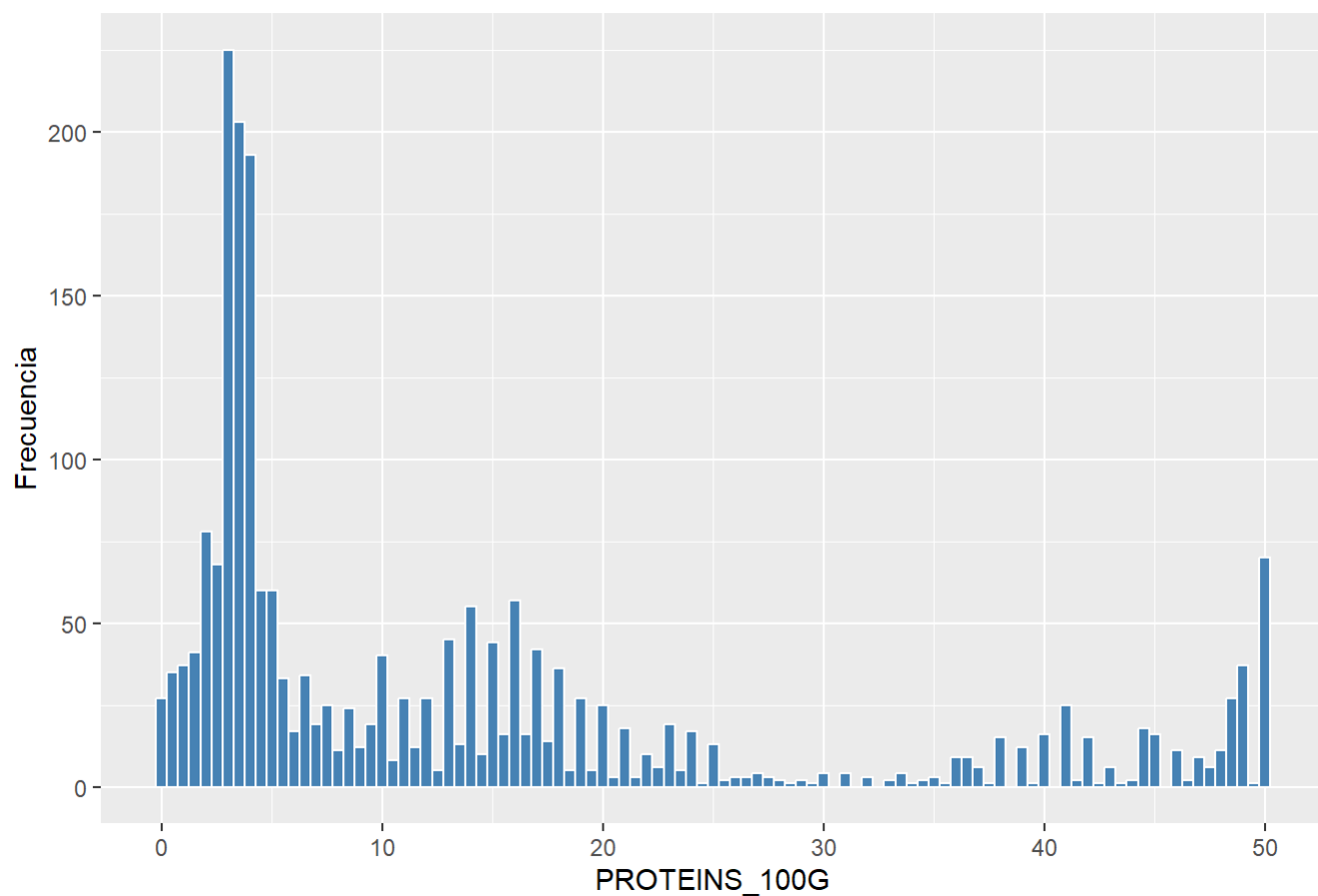
##	PRODUCT_NAME	PROTEINS_100G	CARBOHYDRATES_100G	FAT_100G
##	Length:2219	Min. : 0.00	Min. : 0.00	Min. : 0.000
##	Class :character	1st Qu.: 3.40	1st Qu.: 3.30	1st Qu.: 2.000
##	Mode :character	Median : 6.70	Median : 8.90	Median : 4.800
##		Mean :13.89	Mean :11.65	Mean : 7.118
##		3rd Qu.:18.00	3rd Qu.:15.00	3rd Qu.:10.000
##		Max. :50.00	Max. :56.50	Max. :41.800

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(df_soja, is.numeric)

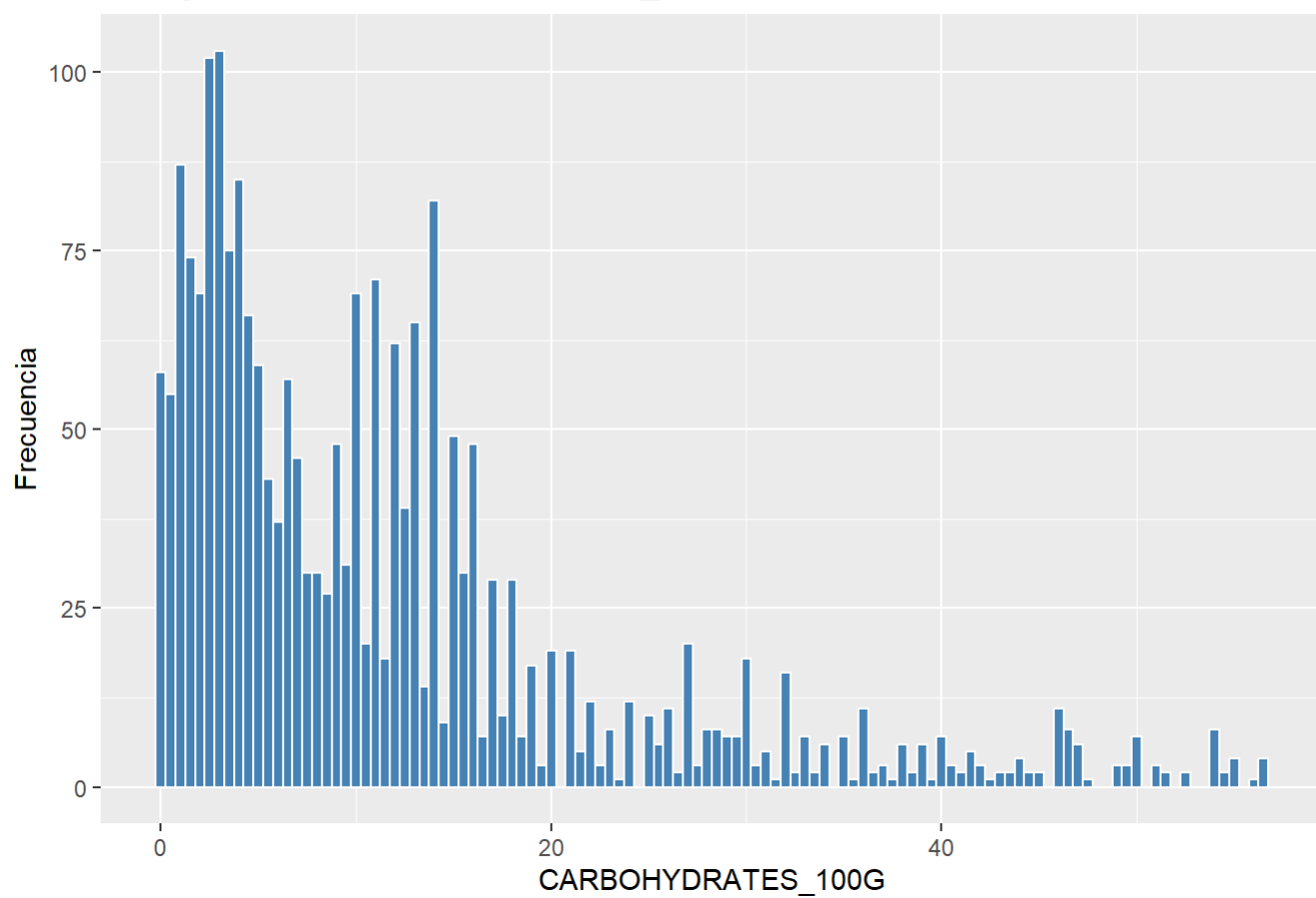
# Crear un histograma para cada columna cuantitativa
for (columna in names(df_soja[columnas_cuantitativas])) {
  plot_data <- df_soja[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

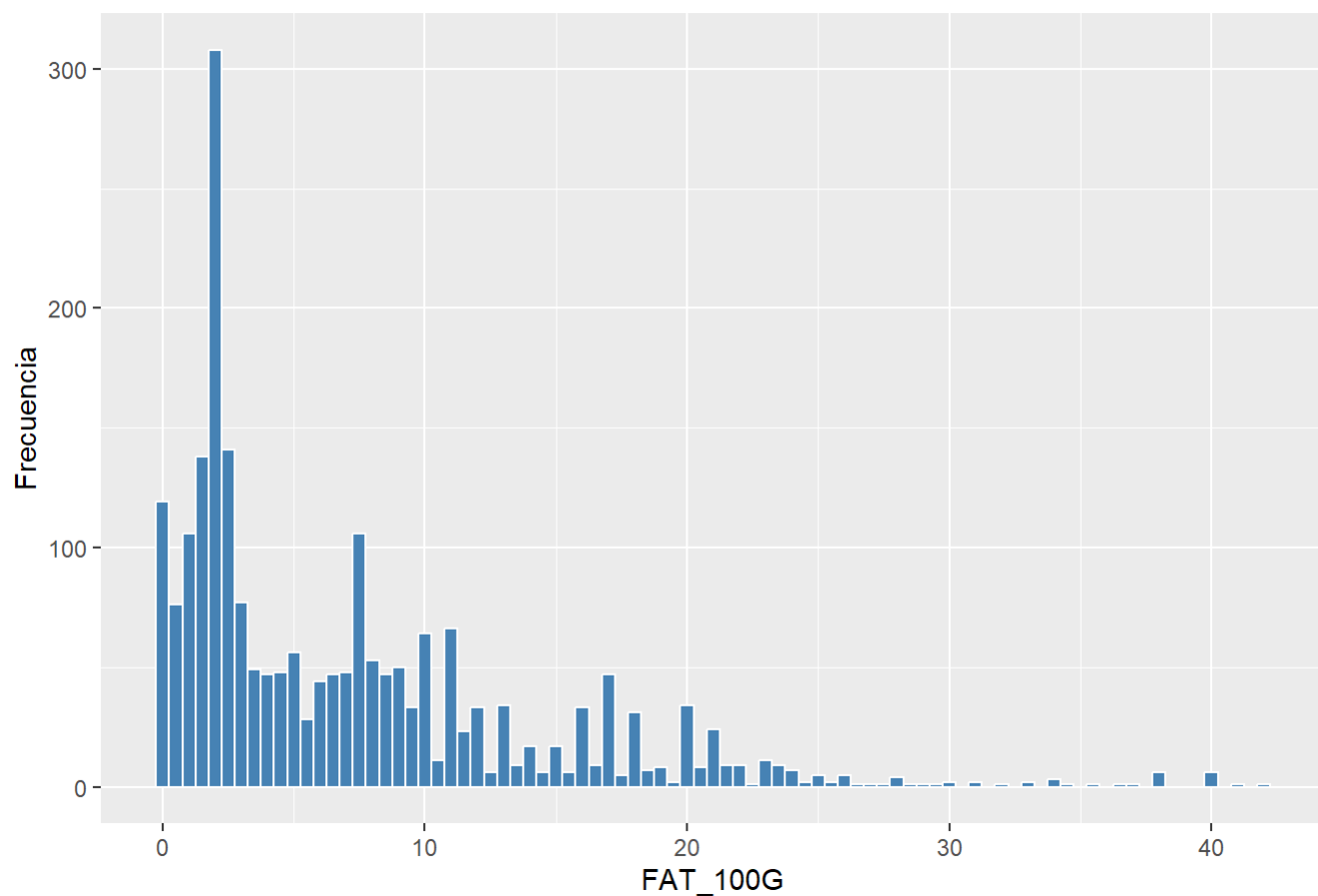
Histograma de PROTEINS_100G



Histograma de CARBOHYDRATES_100G



Histograma de FAT_100G



```
s_soja <- scale(df_soja[, -1])
s_soja1 <- as.data.frame(s_soja)
summary(s_soja)
```

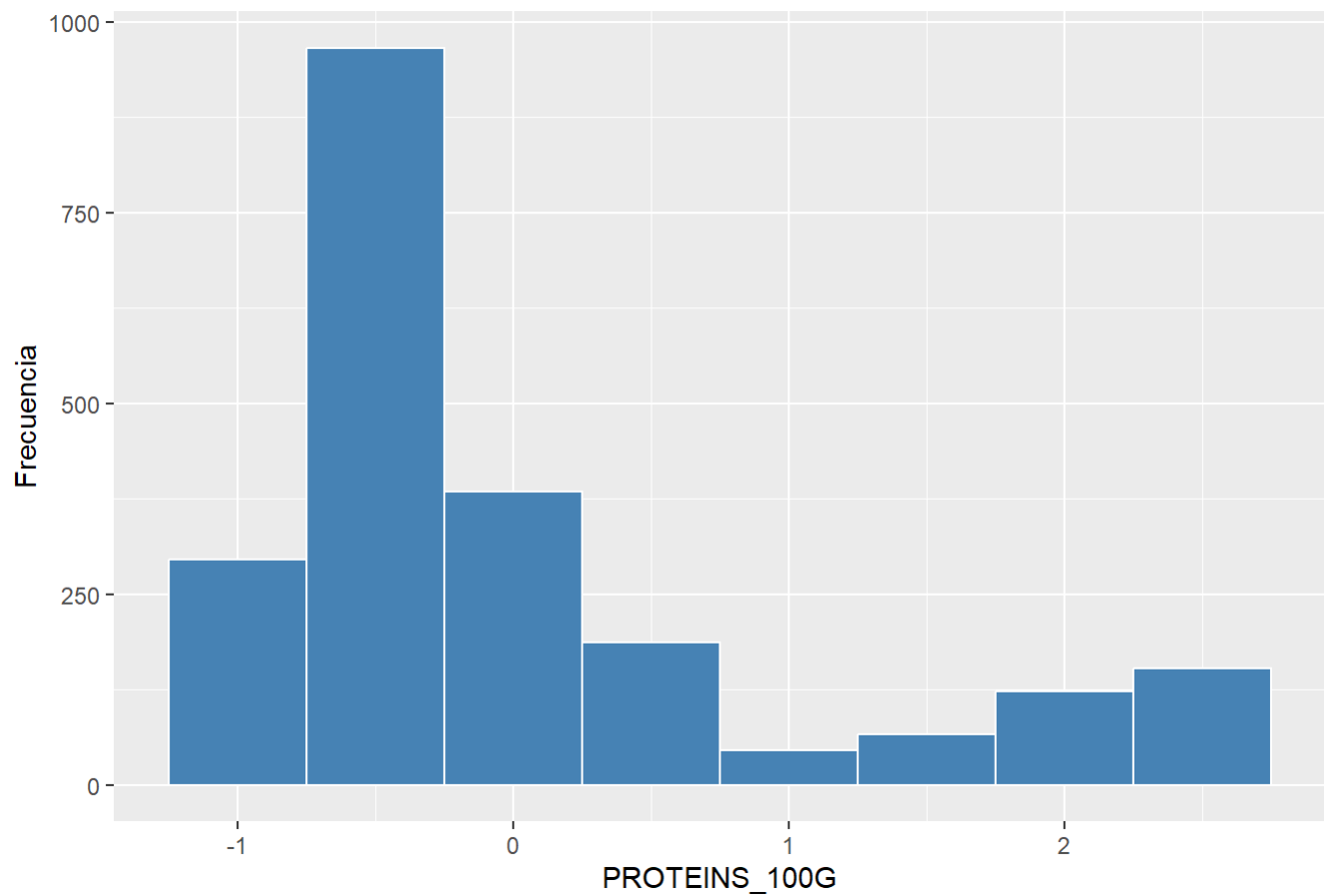
```
## PROTEINS_100G CARBOHYDRATES_100G FAT_100G
## Min. :-0.9430 Min. :-1.0186 Min. :-1.0103
## 1st Qu.: -0.7122 1st Qu.: -0.7300 1st Qu.: -0.7264
## Median : -0.4882 Median : -0.2403 Median : -0.3290
## Mean : 0.0000 Mean : 0.0000 Mean : 0.0000
## 3rd Qu.: 0.2790 3rd Qu.: 0.2931 3rd Qu.: 0.4092
## Max. : 2.4514 Max. : 3.9222 Max. : 4.9231
```

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(s_soja1, is.numeric)

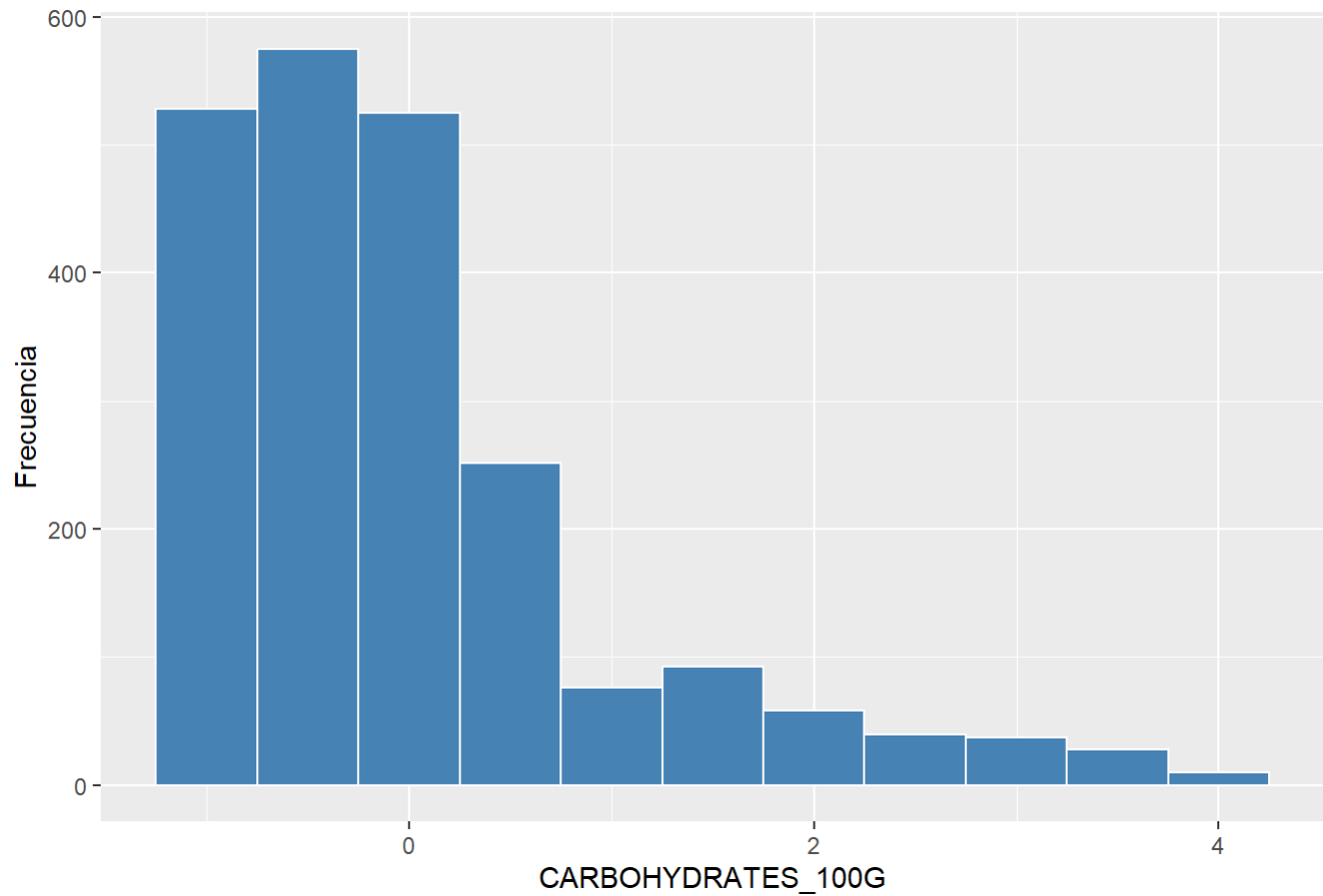
# Crear un histograma para cada columna cuantitativa
for (columna in names(s_soja1[columnas_cuantitativas])) {
  plot_data <- s_soja1[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

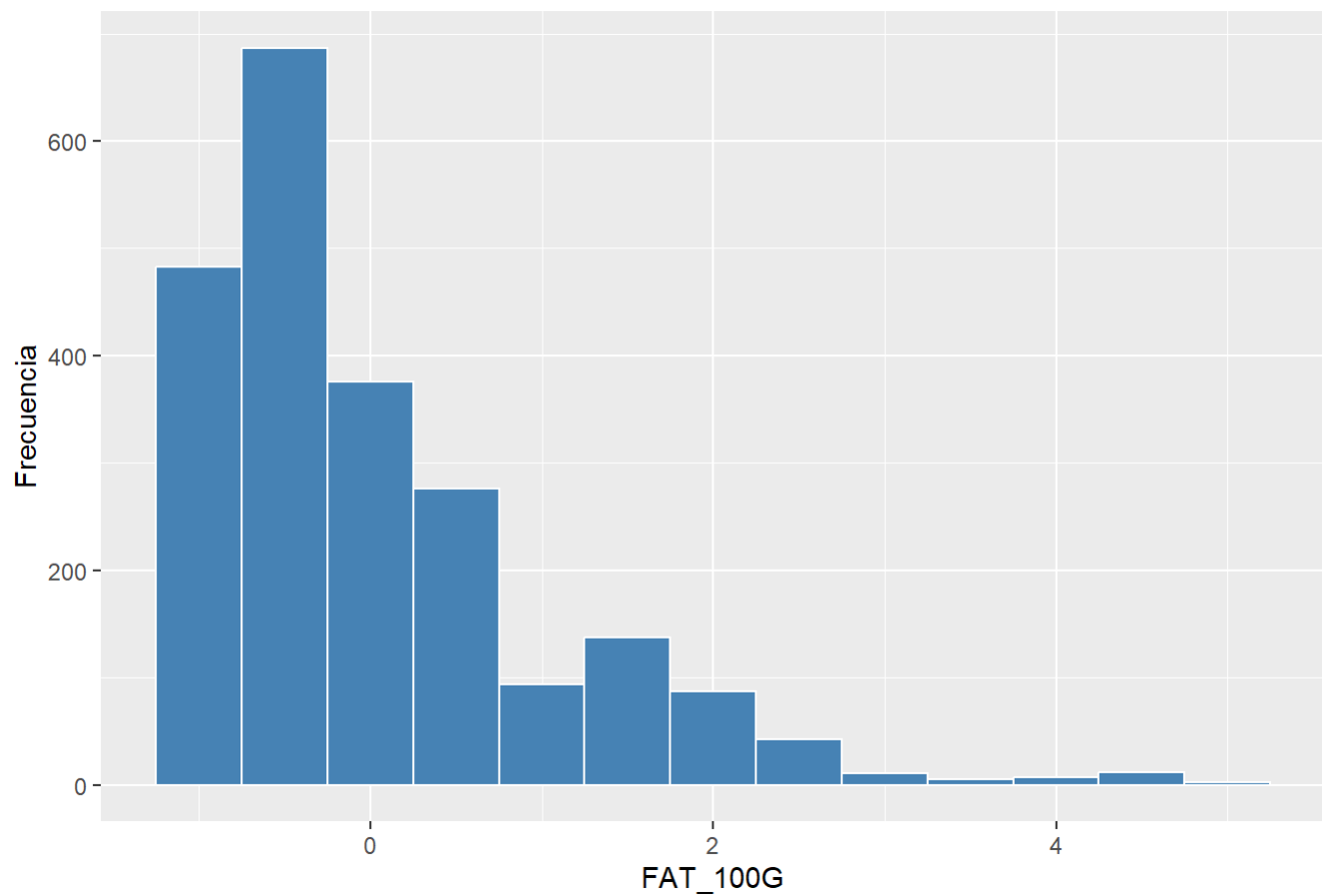

Histograma de PROTEINS_100G



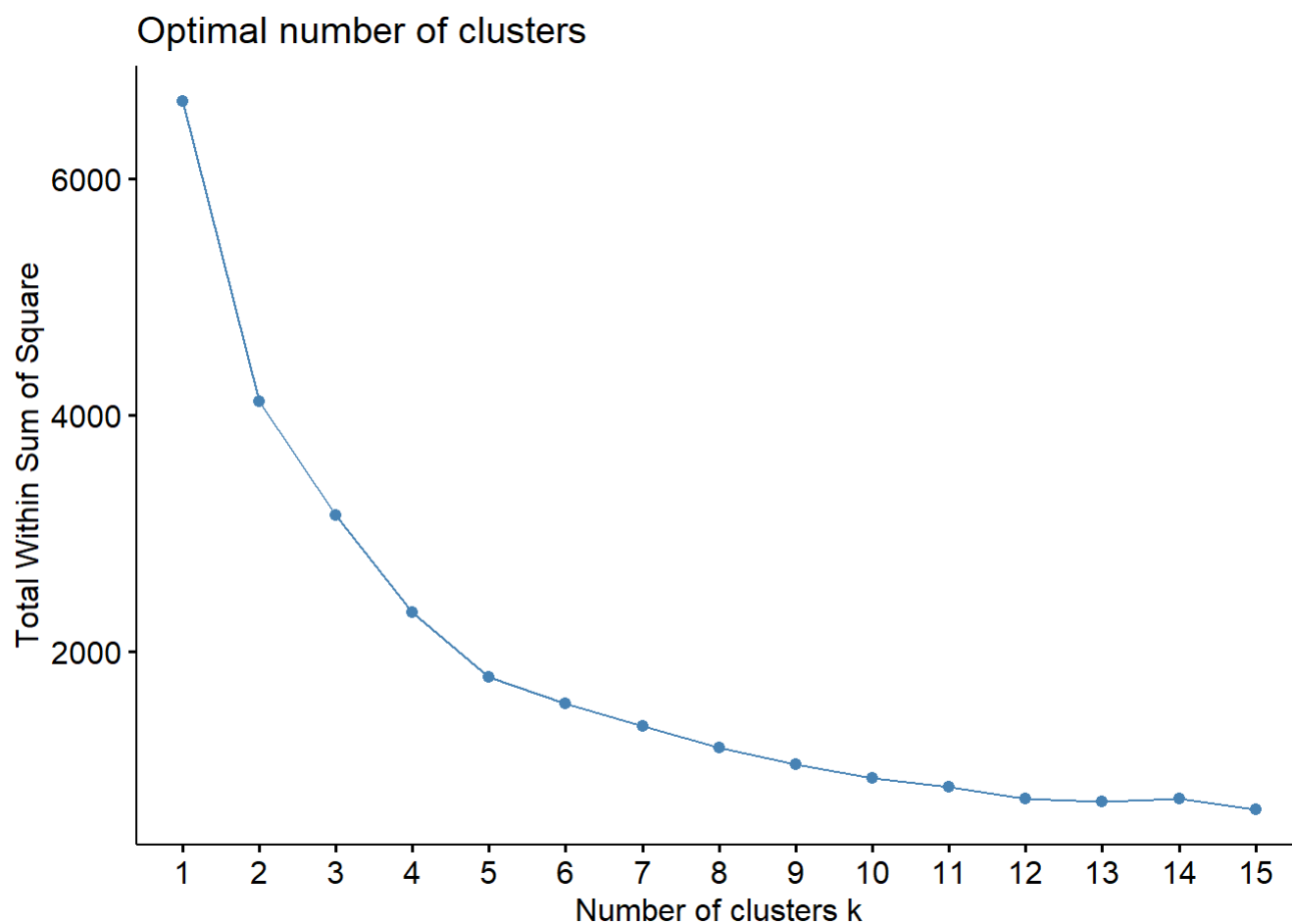
Histograma de CARBOHYDRATES_100G



Histograma de FAT_100G

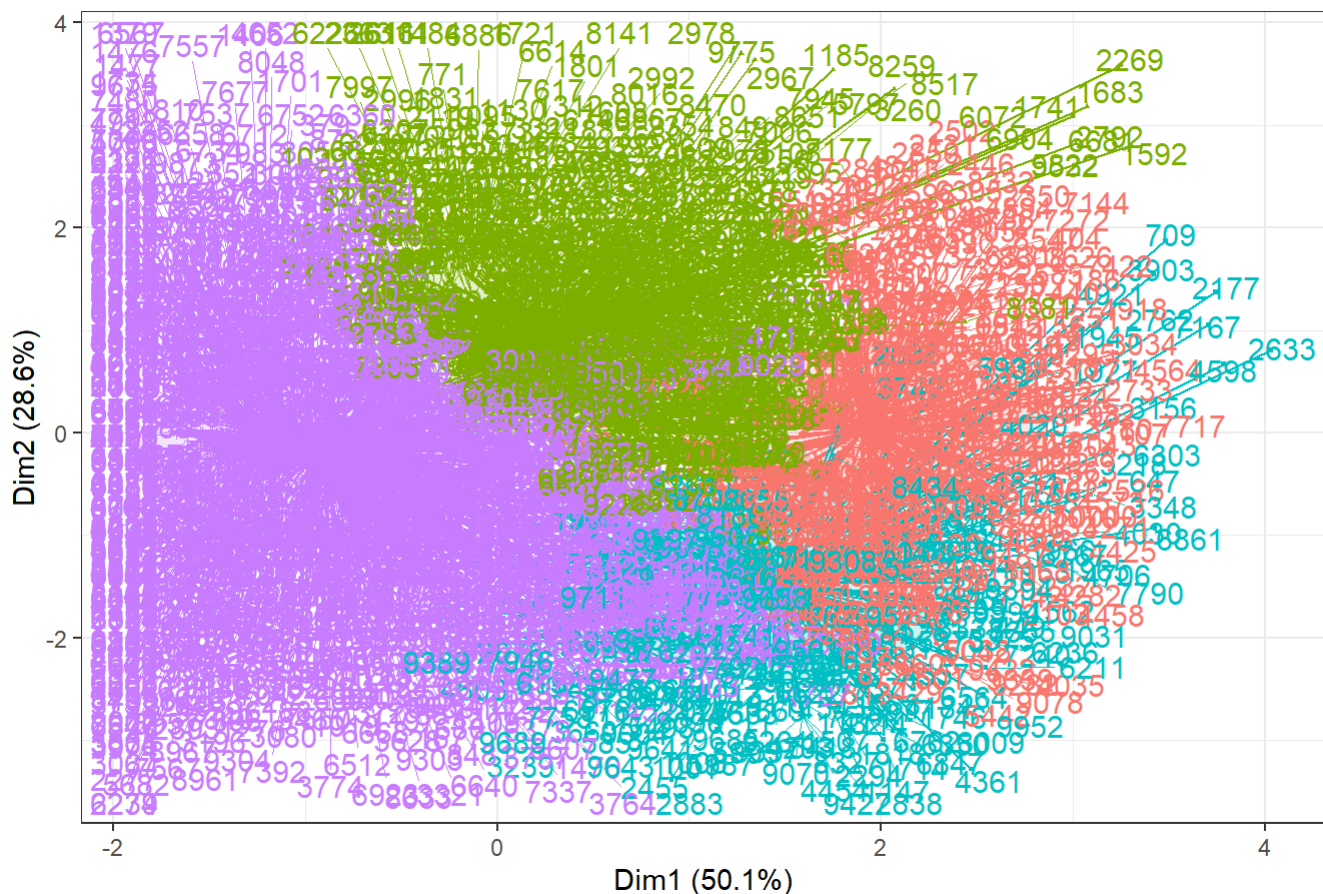


```
#total de cluster óptimos
elbow <- fviz_nbclust(x = s_soja, FUNcluster = kmeans, method = "wss", k.max = 15,
                      diss = get_dist(s_soja, method = "euclidean"), nstart = 25)
print(elbow)
```



```
#Clustering K-means
set.seed(123)
km_clusters <- kmeans(x=s_soja,centers=4,nstart=25)
fviz_cluster(object=km_clusters,data=s_soja,show.clust.cent = TRUE,
              ellipse.type="euclid",star.plot=TRUE,repel=TRUE,
              pointsize=0.5,outlier.color="darkred") +
  labs(title ="Resultados clustering K-means") +
  theme_bw() +
  theme(legend.position = "none")
```

Resultados clustering K-means

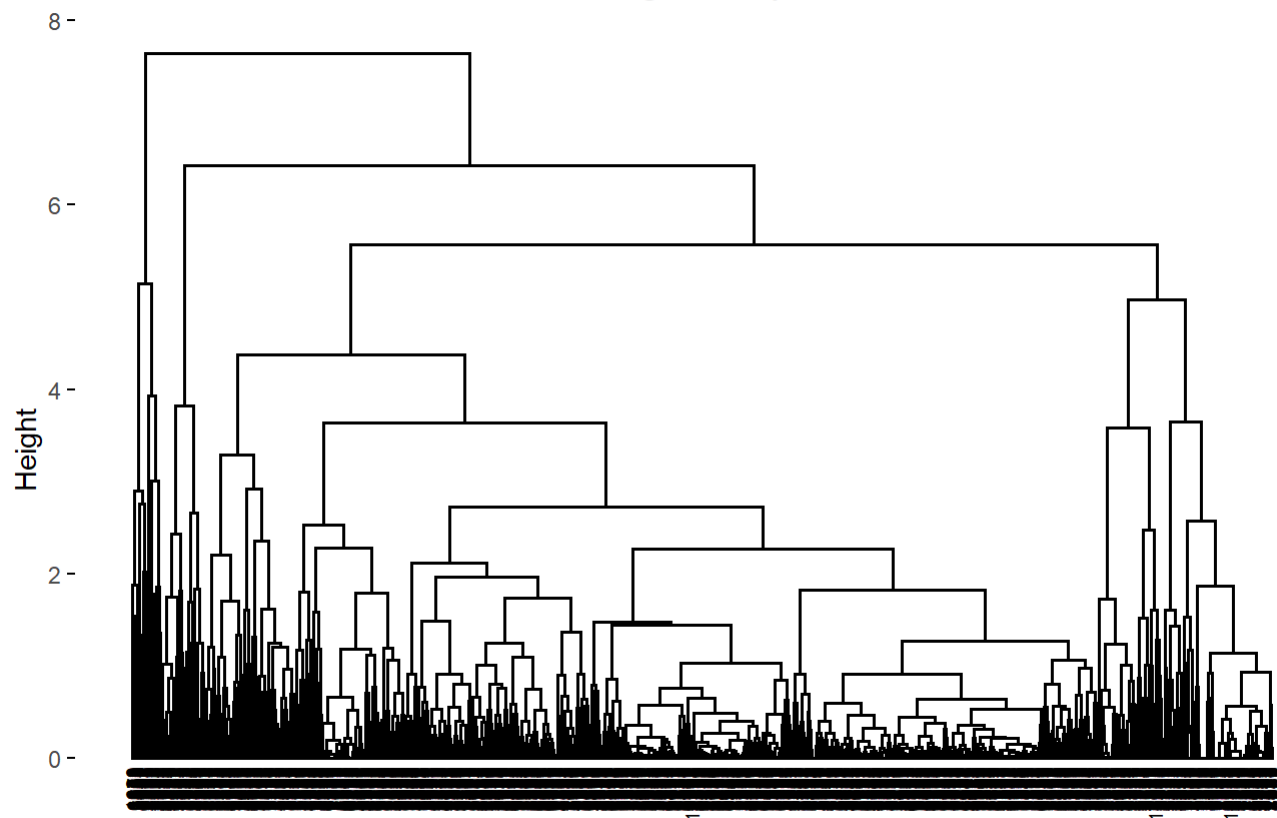


```
#Dedrograma
set.seed(101)
heculidea <- hclust(d = dist(x = s_soja, method = "euclidean"),
  method = "complete")

fviz_dend(x=heculidea,cex=0.5,main = "Linkage completo",
  sub="Distancia euclídea")+
  theme(plot.title = element_text(hjust=0.5,size=15))
```

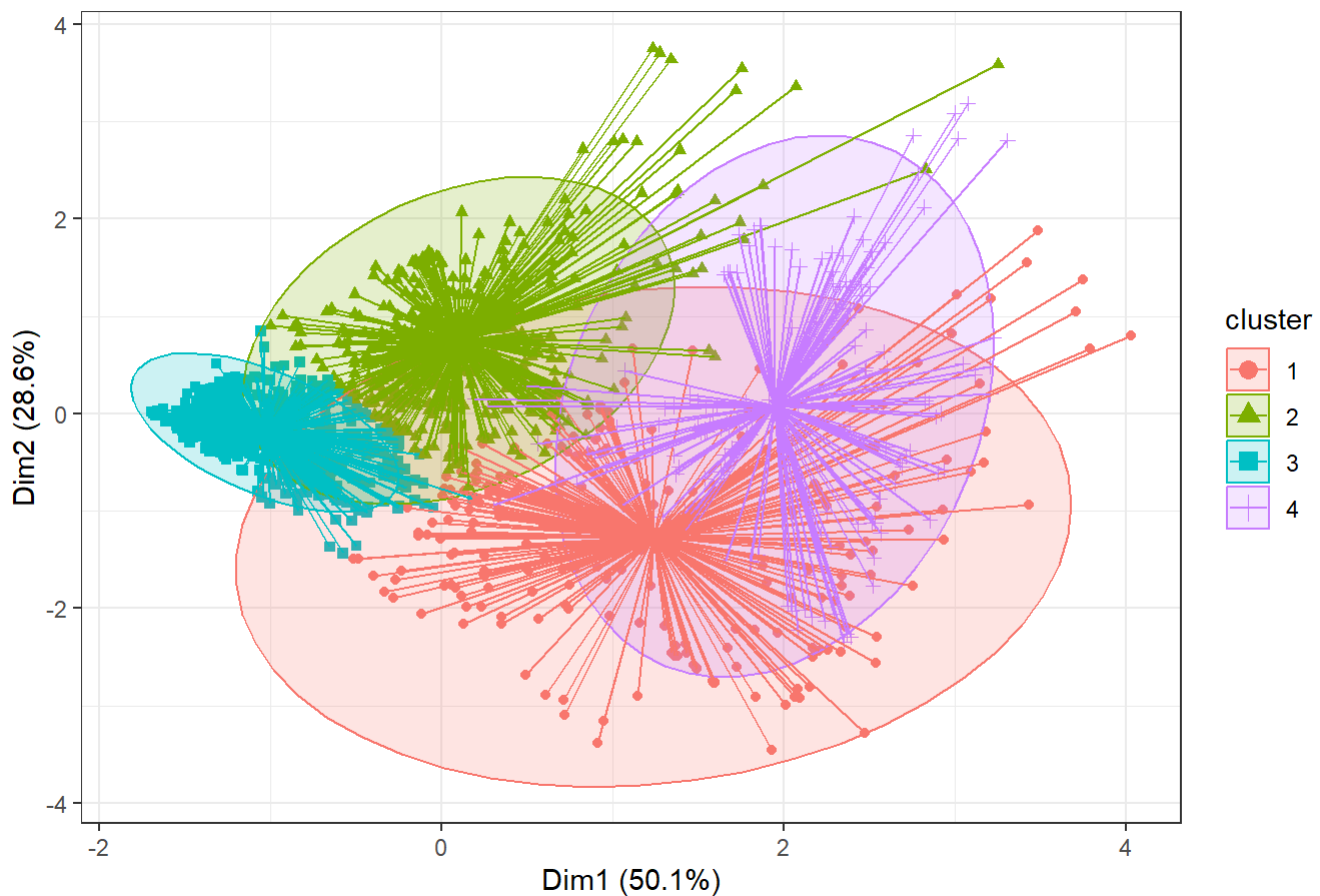
```
## Warning: The `scale` argument of `guides()` cannot be `FALSE`. Use "none" instead as
## of ggplot2 3.3.4.
## i The deprecated feature was likely used in the factoextra package.
## Please report the issue at <https://github.com/kassambara/factoextra/issues>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

Linkage completo



```
#Resultados clustering K-means
require(cluster)
pam.res <- pam(s_soja, 4)
# Visualización
fviz_cluster(pam.res, geom = "point", ellipse.type = "norm",
              show.clust.cent = TRUE, star.plot = TRUE)+
  labs(title = "Resultados clustering K-means")+ theme_bw()
```

Resultados clustering K-means



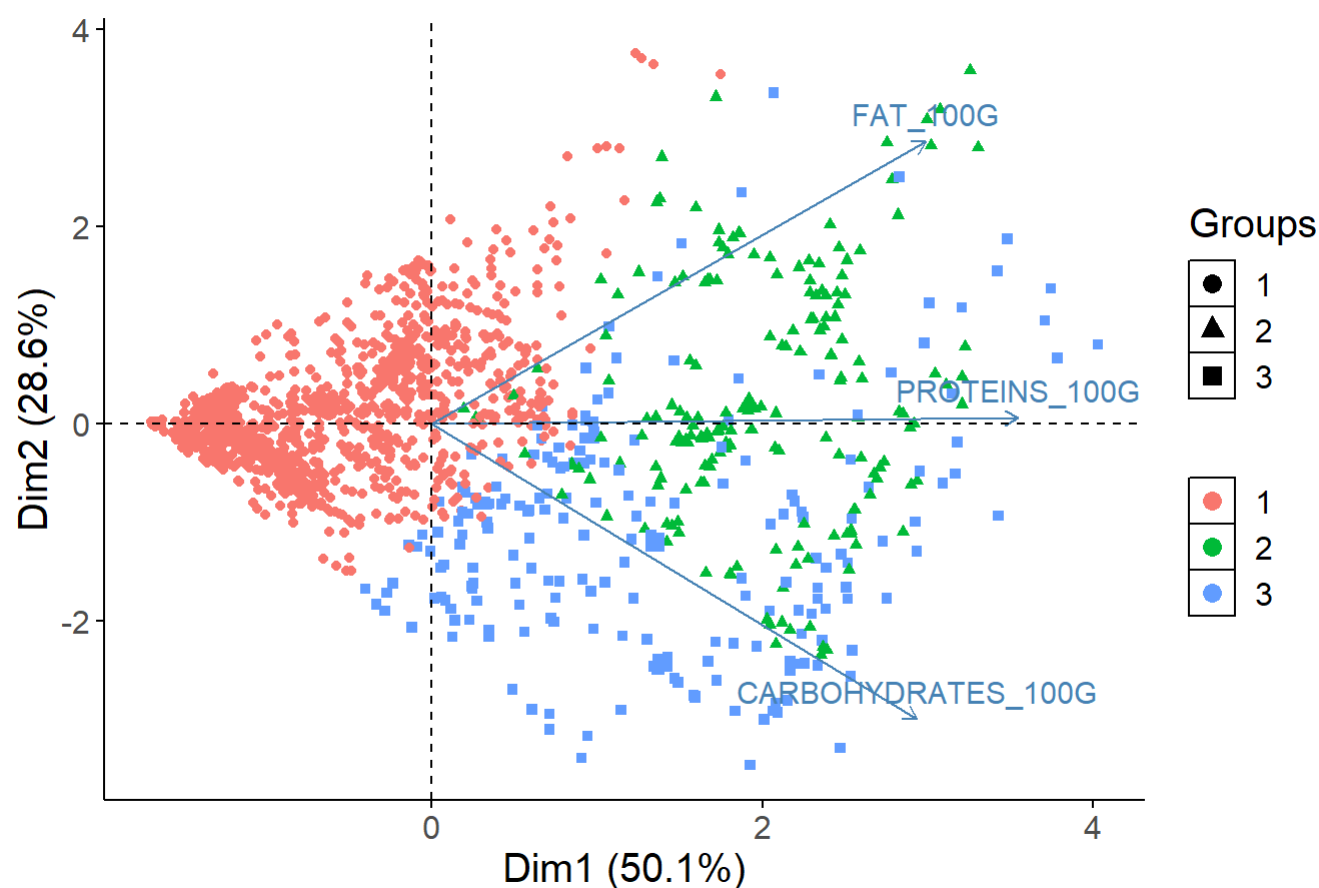
Con esta imagen se evidencia que hay cierto grado de conflicto entre los clusters, ya que tienen cierta área solapada entre ambos. Estas representaciones mezclan cluster y PCA, pero no indican el grado de representación de cada variable en cada uno de los componentes.

Biplot PCA y K-Means para medir representatividad soja

```
# PCA
pca <- prcomp(df_soja[, -1], scale=TRUE)
df_soja.pca <- pca$x

# Cluster over the three first PCA dimensions
kc <- kmeans(df_soja.pca[, 1:3], 3)

fviz_pca_biplot(pca, label="var", habillage=as.factor(kc$cluster)) +
  labs(color=NULL) + ggtitle("") +
  theme(text = element_text(size = 15),
        panel.background = element_blank(),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        axis.line = element_line(colour = "black"),
        legend.key = element_rect(fill = "white"))
```



SEITAN

```
summary(df_seitan)
```

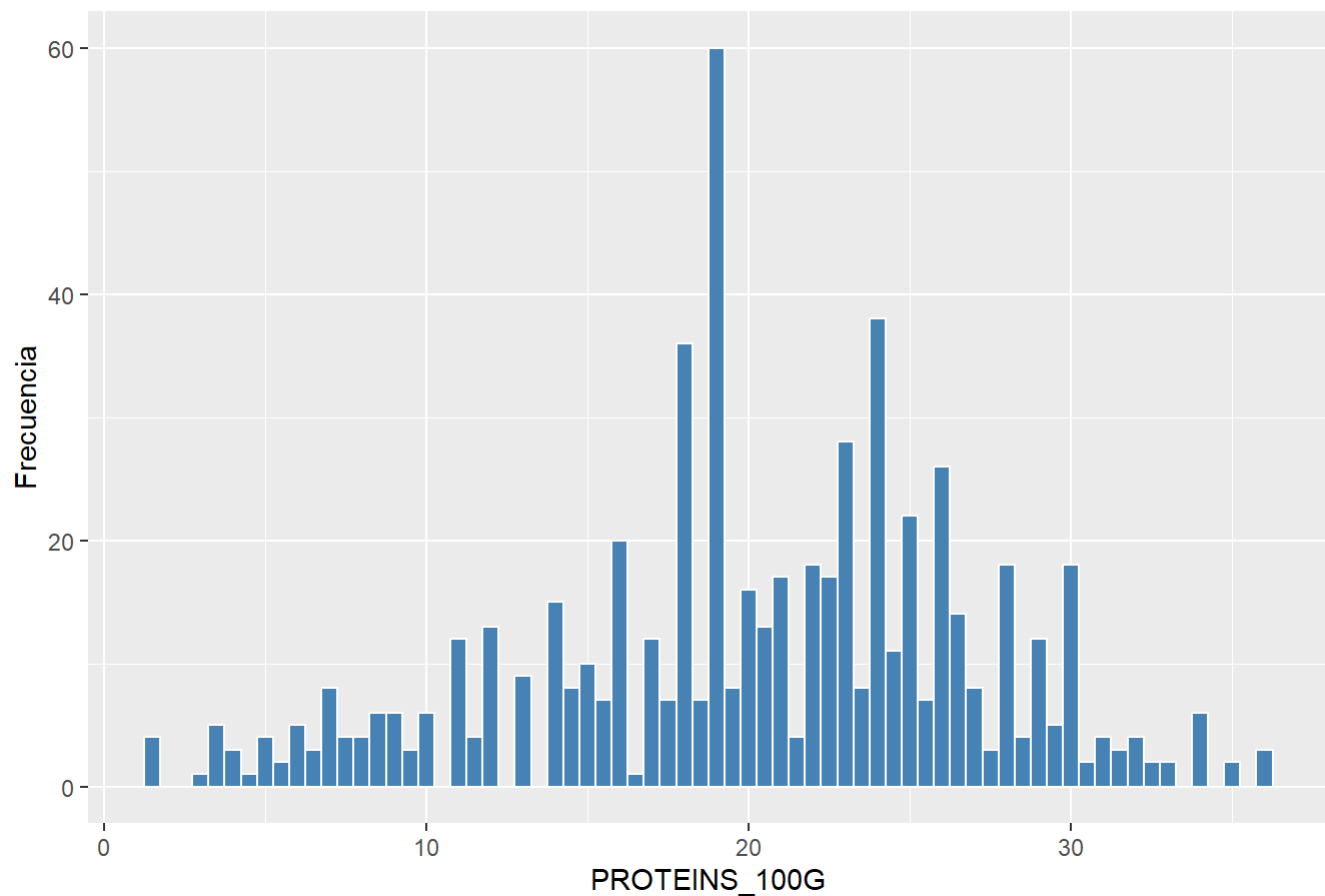
##	PRODUCT_NAME	PROTEINS_100G	CARBOHYDRATES_100G	FAT_100G
##	Length:629	Min. : 1.40	Min. : 0.000	Min. : 0.000
##	Class :character	1st Qu.:16.00	1st Qu.: 5.200	1st Qu.: 1.600
##	Mode :character	Median :20.49	Median : 7.000	Median : 3.800
##		Mean :20.15	Mean : 8.419	Mean : 5.194
##		3rd Qu.:25.00	3rd Qu.:11.000	3rd Qu.: 8.000
##		Max. :36.00	Max. :24.400	Max. :17.400

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(df_seitan, is.numeric)

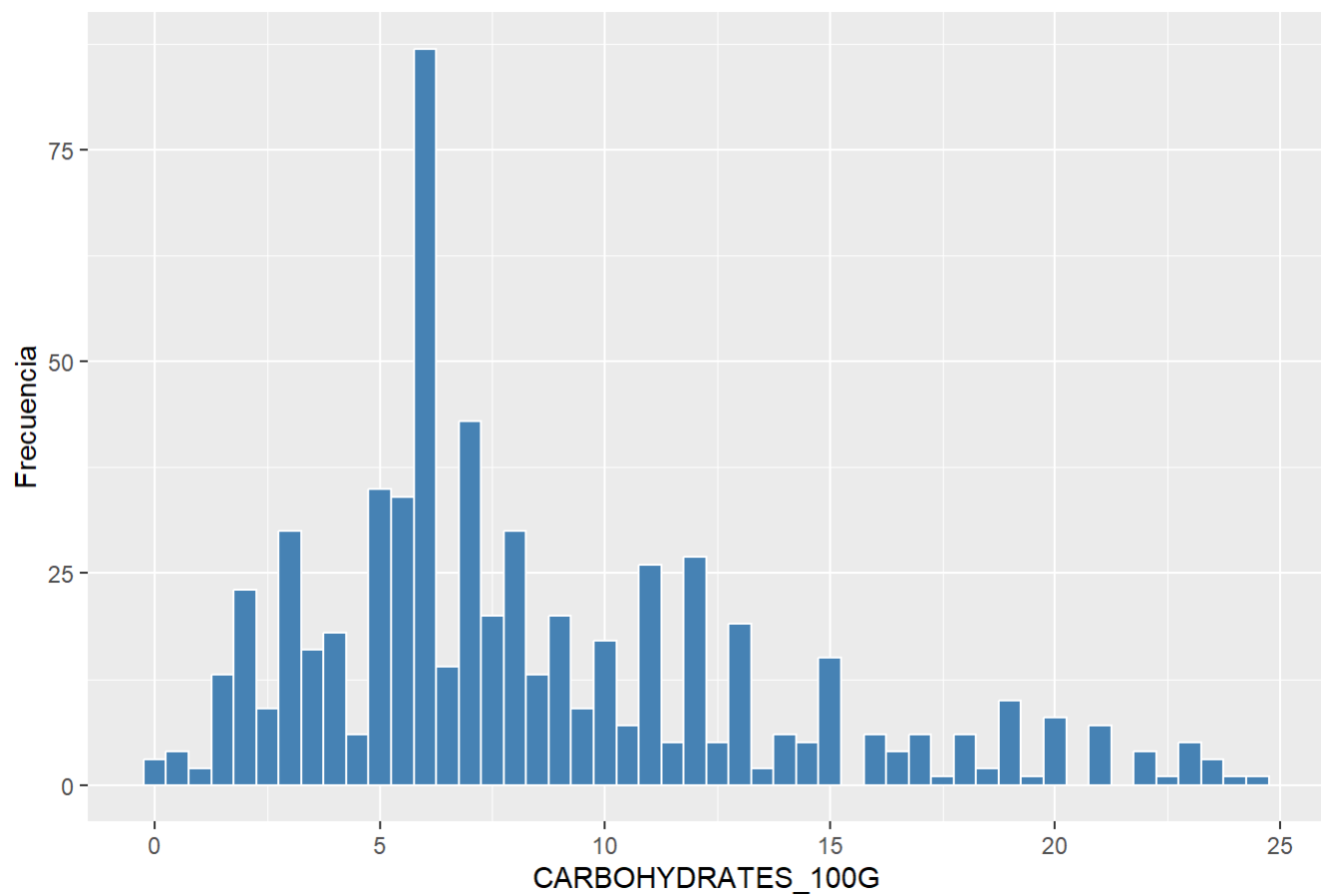
# Crear un histograma para cada columna cuantitativa
for (columna in names(df_seitan[columnas_cuantitativas])) {
  plot_data <- df_seitan[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

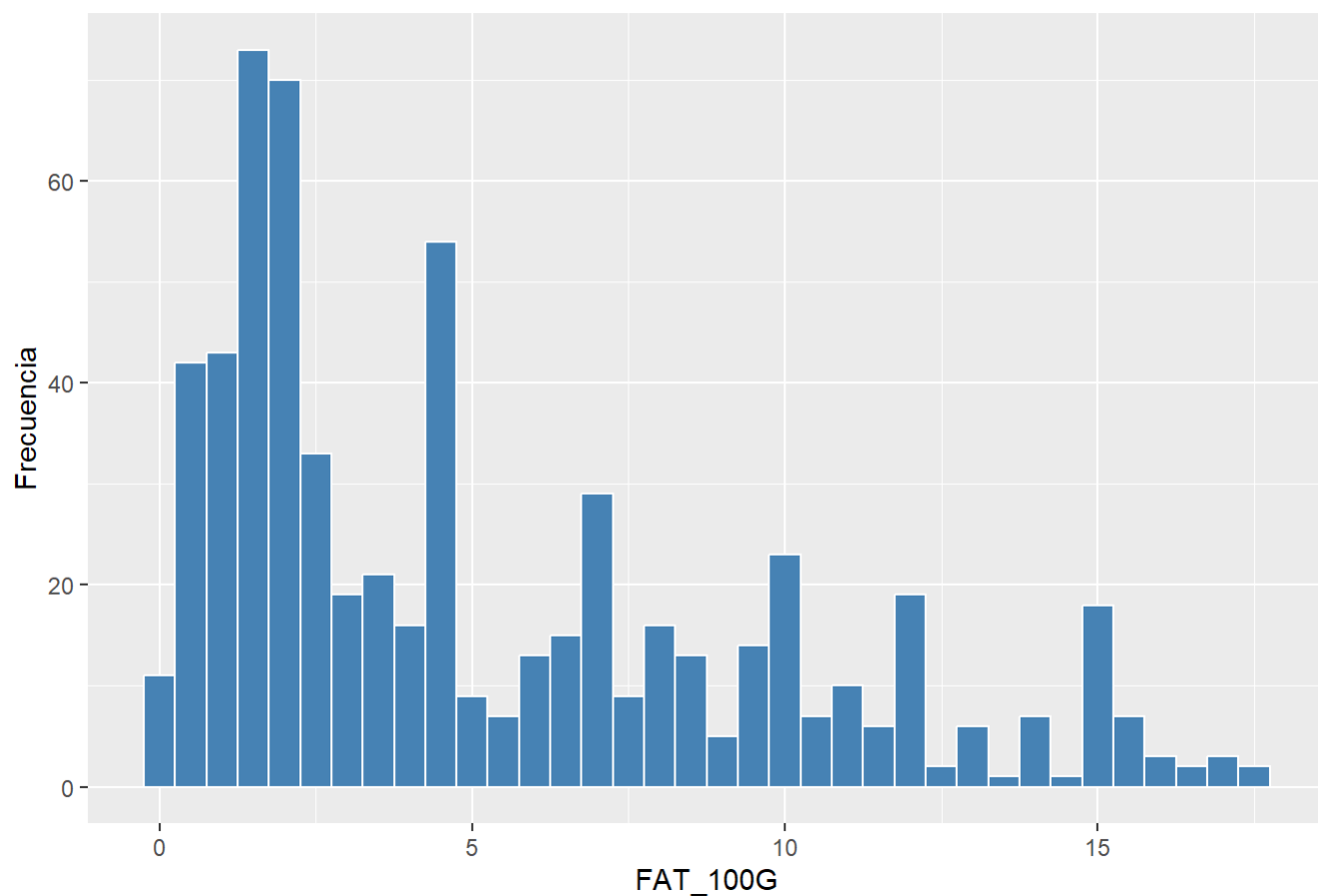

Histograma de PROTEINS_100G



Histograma de CARBOHYDRATES_100G



Histograma de FAT_100G



```
s_seitan <- scale(df_seitan[, -1])
s_seitan1 <- as.data.frame(s_seitan)
summary(s_seitan)
```

```
## PROTEINS_100G CARBOHYDRATES_100G FAT_100G
## Min. :-2.7208 Min. :-1.6559 Min. :-1.1841
## 1st Qu.: -0.6028 1st Qu.: -0.6331 1st Qu.: -0.8193
## Median : 0.0486 Median : -0.2790 Median : -0.3178
## Mean : 0.0000 Mean : 0.0000 Mean : 0.0000
## 3rd Qu.: 0.7029 3rd Qu.: 0.5078 3rd Qu.: 0.6397
## Max. : 2.2987 Max. : 3.1436 Max. : 2.7827
```

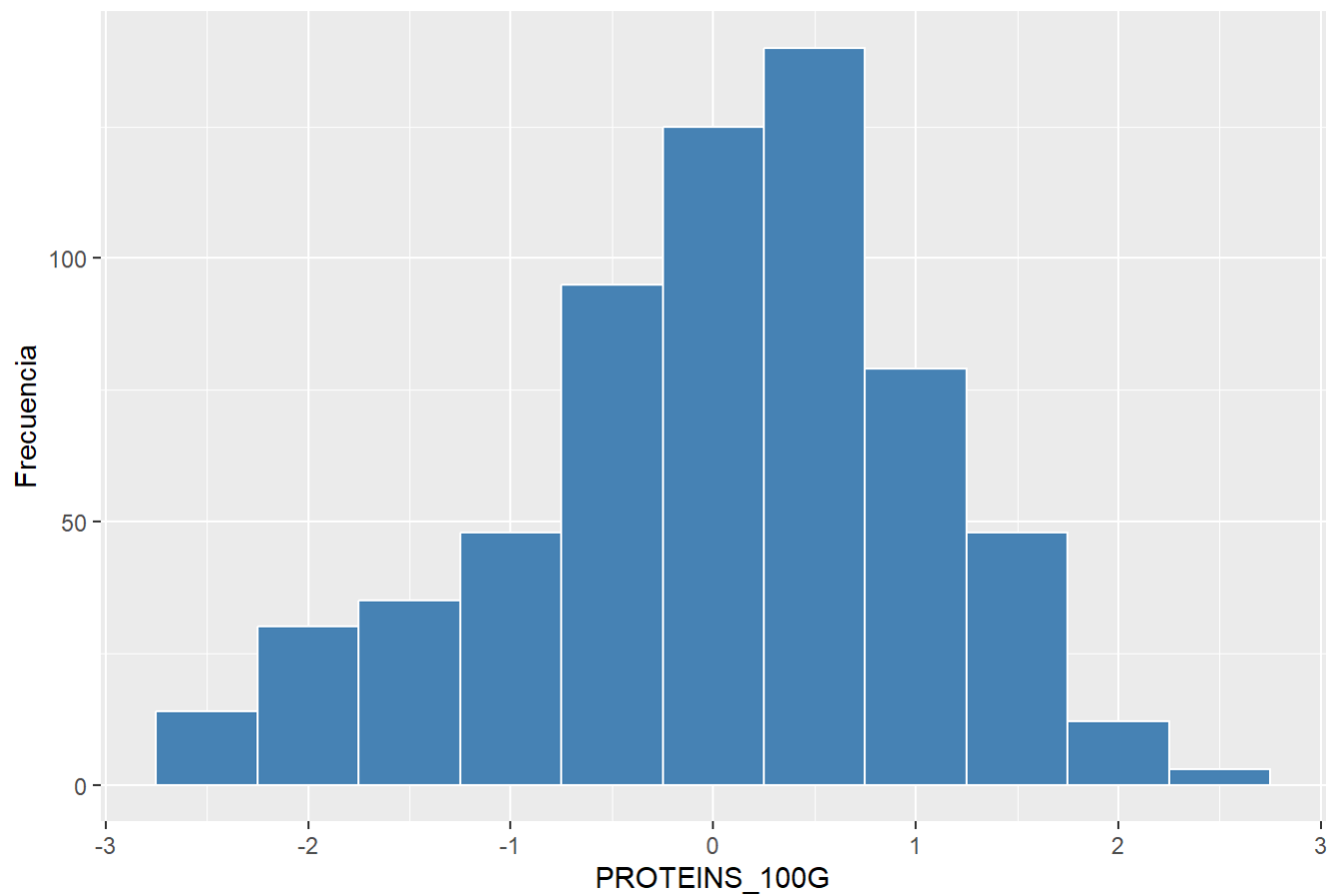
#Datos normalizados para el Seitán

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(s_seitan1, is.numeric)

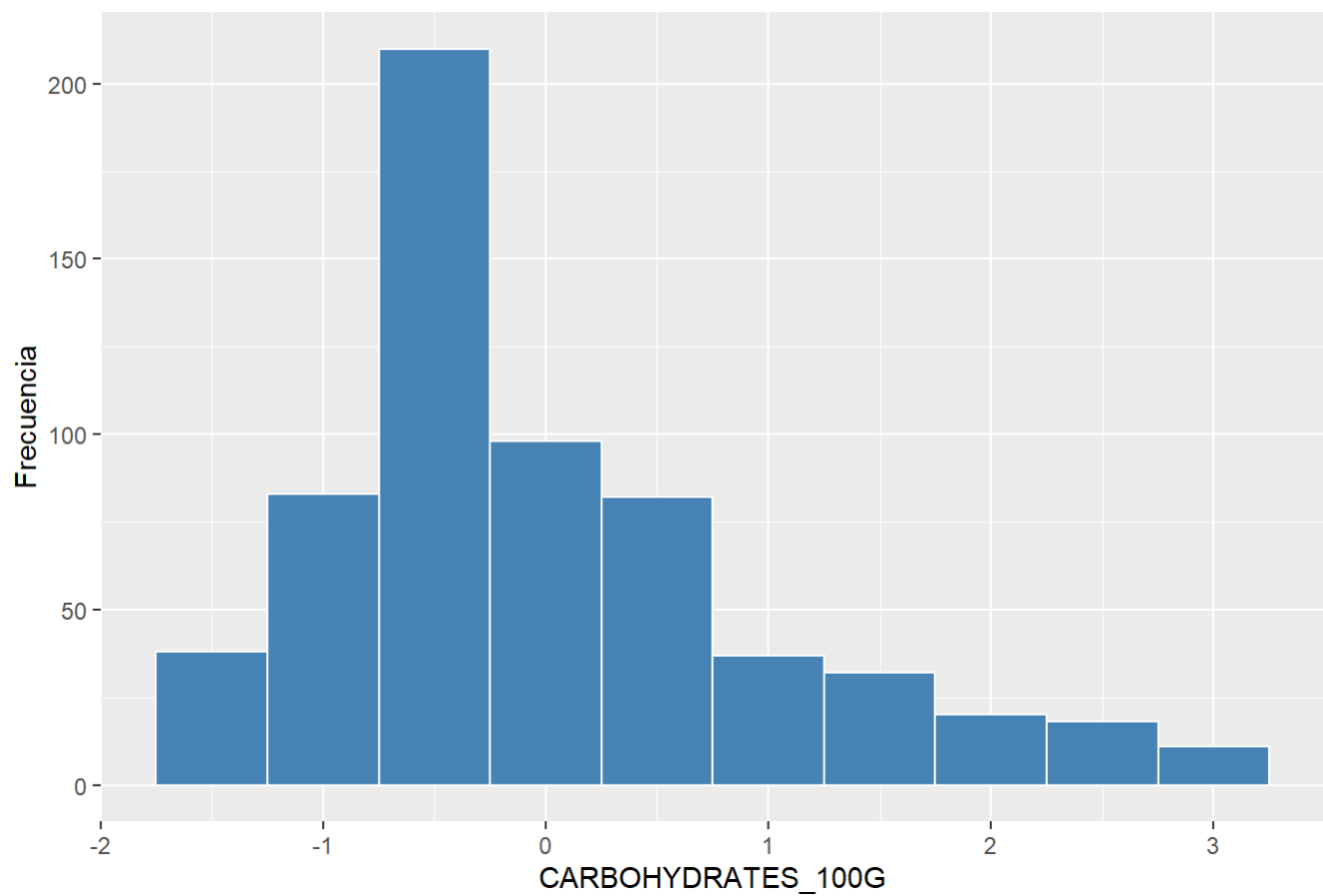
# Crear un histograma para cada columna cuantitativa
for (columna in names(s_seitan1[columnas_cuantitativas])) {
  plot_data <- s_seitan1[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

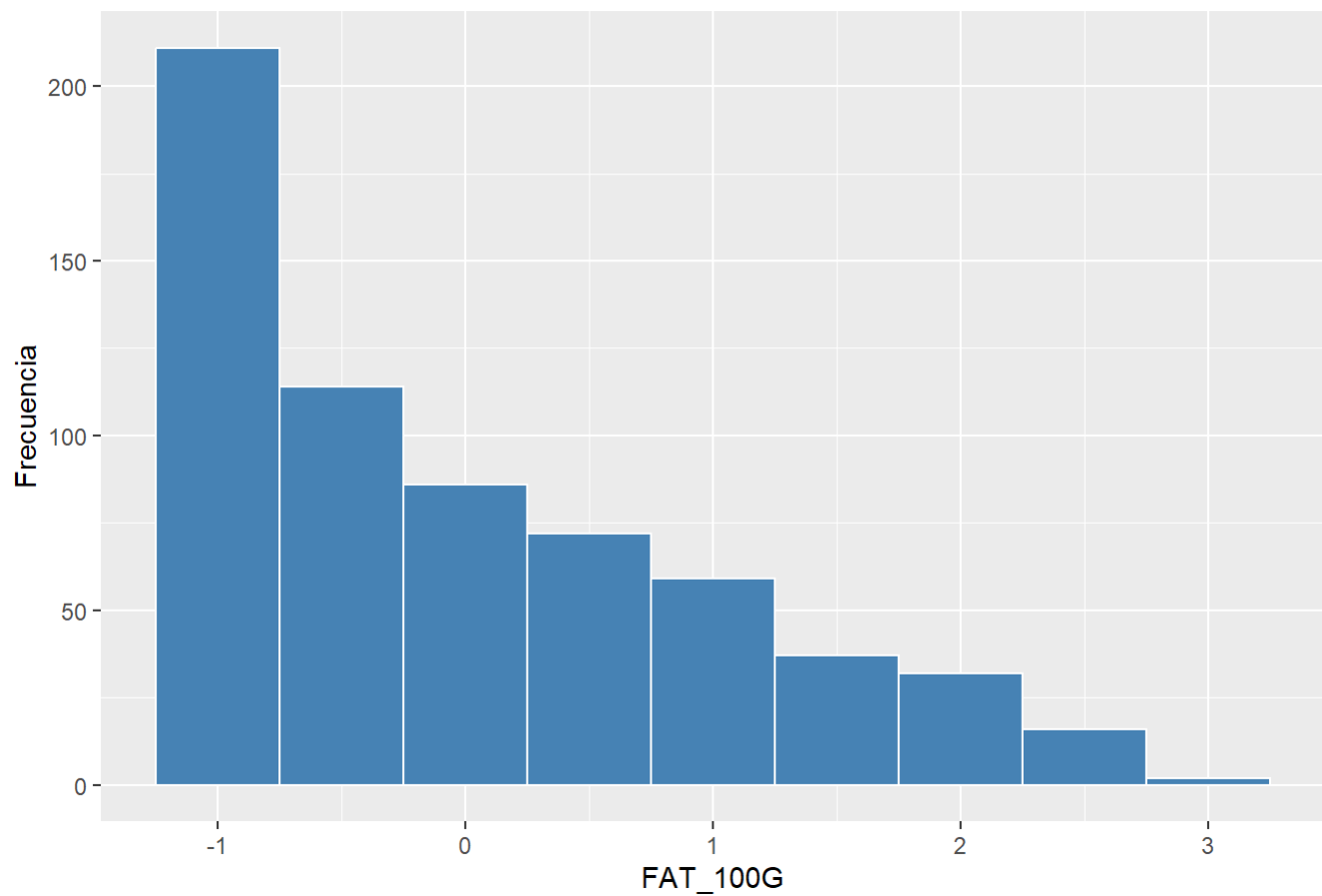
Histograma de PROTEINS_100G



Histograma de CARBOHYDRATES_100G



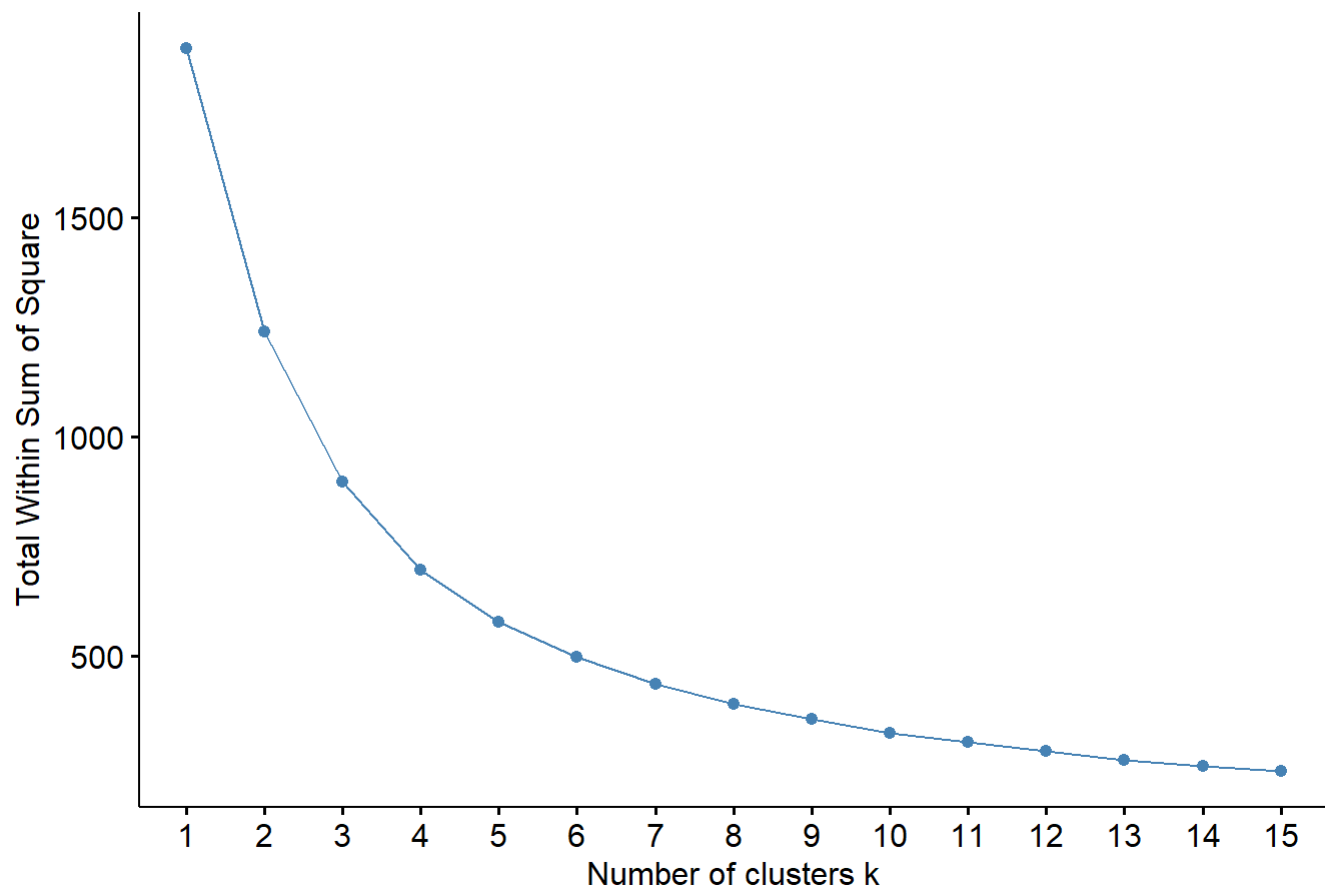
Histograma de FAT_100G



```
s_seitan <- scale(df_seitan[,-1])

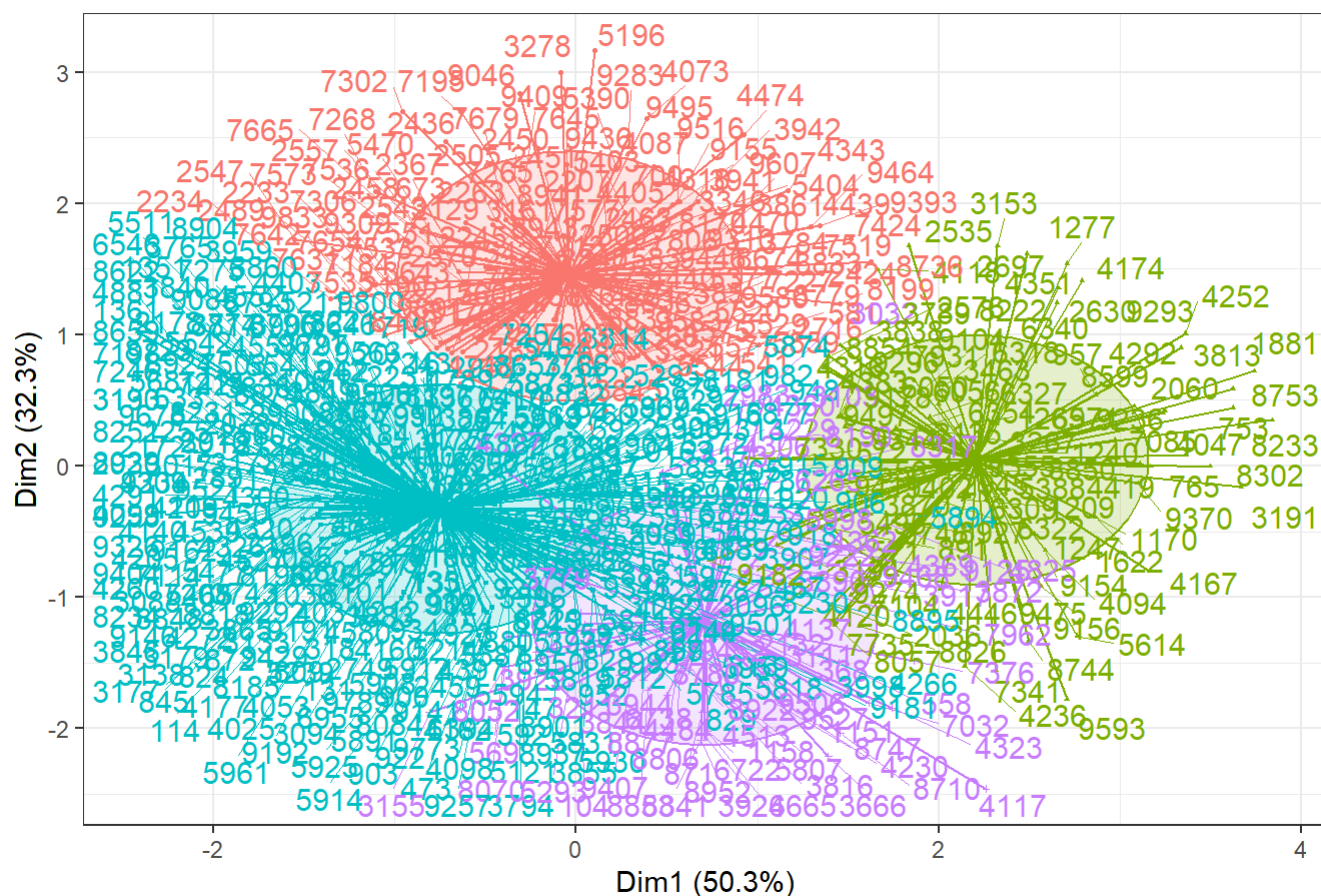
# total de cluster óptimos
elbow <- fviz_nbclust(x = s_seitan, FUNcluster = kmeans, method = "wss", k.max = 15,
                      diss = get_dist(s_seitan, method = "euclidean"), nstart = 25)
print(elbow)
```

Optimal number of clusters



```
set.seed(123)
km_clusters <- kmeans(x=s_seitan,centers=4,nstart=25)
fviz_cluster(object=km_clusters,data=s_seitan,show.clust.cent = TRUE,
              ellipse.type="euclid",star.plot=TRUE,repel=TRUE,
              pointsize=0.5,outlier.color="darkred") +
  labs(title = "Resultados clustering K-means") +
  theme_bw() +
  theme(legend.position = "none")
```

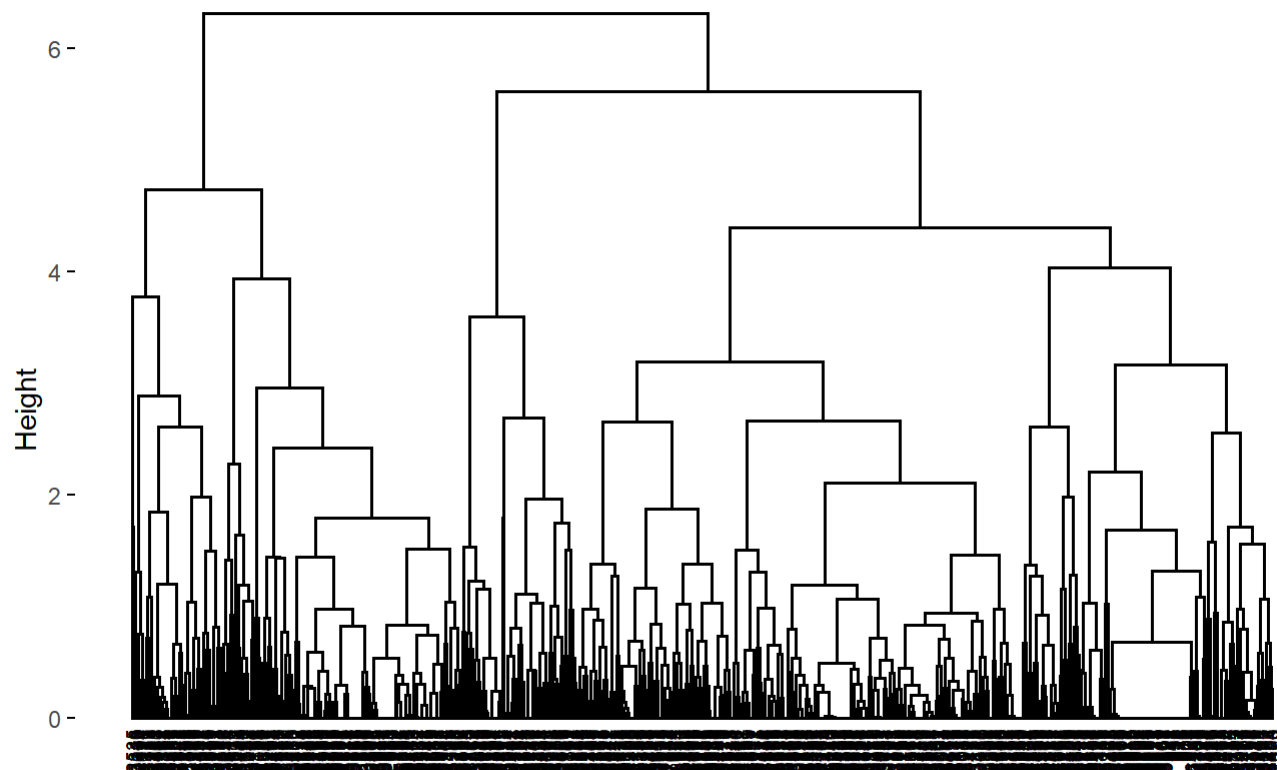
Resultados clustering K-means



```
set.seed(101)
heculidea <- hclust(d = dist(x = s_seitan, method = "euclidean"),
  method = "complete")

fviz_dend(x=heculidea,cex=0.5,main = "Linkage completo",
  sub="Distancia euclídea")+
  theme(plot.title = element_text(hjust=0.5,size=15))
```

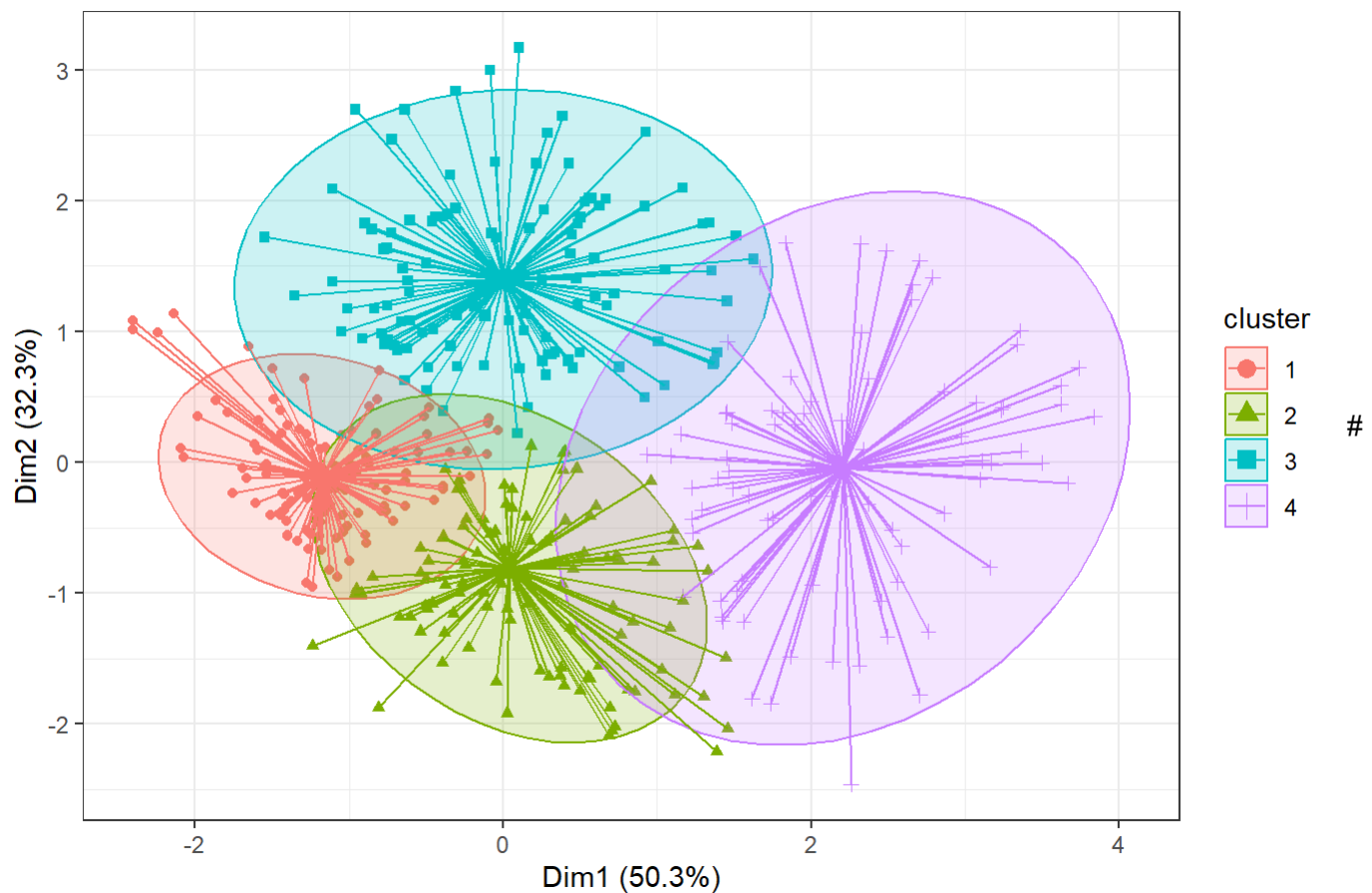
Linkage completo



```
require(cluster)

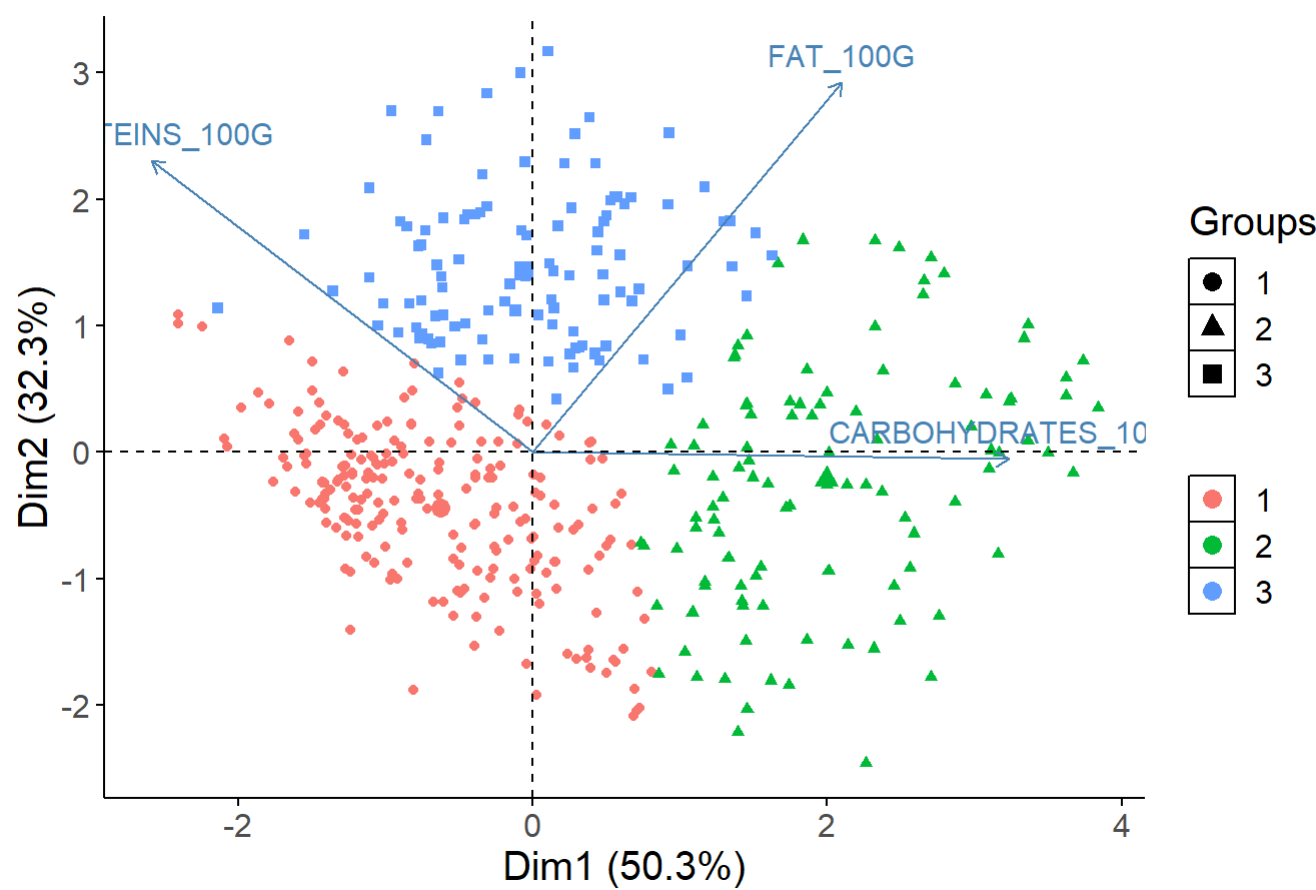
pam.res <- pam(s_seitan, 4)
# Visualización
fviz_cluster(pam.res, geom = "point", ellipse.type = "norm",
              show.clust.cent = TRUE, star.plot = TRUE)+
  labs(title = "Resultados clustering K-means")+ theme_bw()
```


Resultados clustering K-means



Biplot PCA y K-Means para medir representatividad seitan

```
# PCA
pca <- prcomp(df_seitan[, -1], scale=TRUE)
df_seitan.pca <- pca$x
# Cluster over the three first PCA dimensions
kc <- kmeans(df_seitan.pca[, 1:3], 3)
fviz_pca_biplot(pca, label="var", habillage=as.factor(kc$cluster)) +
  labs(color=NULL) + ggtitle("") +
  theme(text = element_text(size = 15),
        panel.background = element_blank(),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        axis.line = element_line(colour = "black"),
        legend.key = element_rect(fill = "white"))
```



TOfu

```
summary(df_tofu)
```

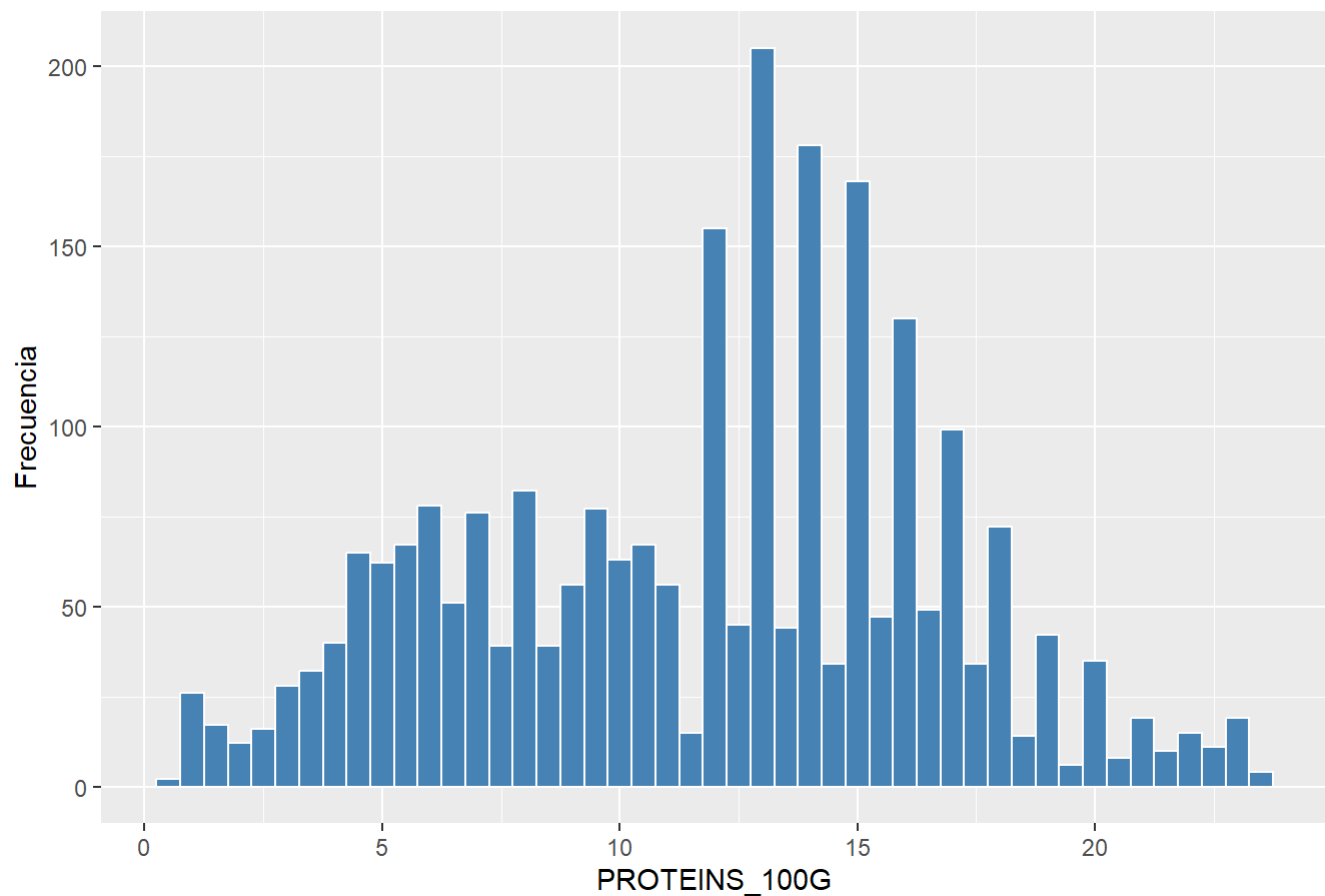
##	PRODUCT_NAME	PROTEINS_100G	CARBOHYDRATES_100G	FAT_100G
##	Length:2509	Min. : 0.50	Min. : 0.000	Min. : 0.00
##	Class :character	1st Qu.: 7.90	1st Qu.: 1.300	1st Qu.: 5.00
##	Mode :character	Median :12.60	Median : 2.400	Median : 7.80
##		Mean :11.78	Mean : 5.414	Mean : 8.18
##		3rd Qu.:15.20	3rd Qu.: 7.500	3rd Qu.:10.30
##		Max. :23.60	Max. :27.429	Max. :21.80

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(df_tofu, is.numeric)

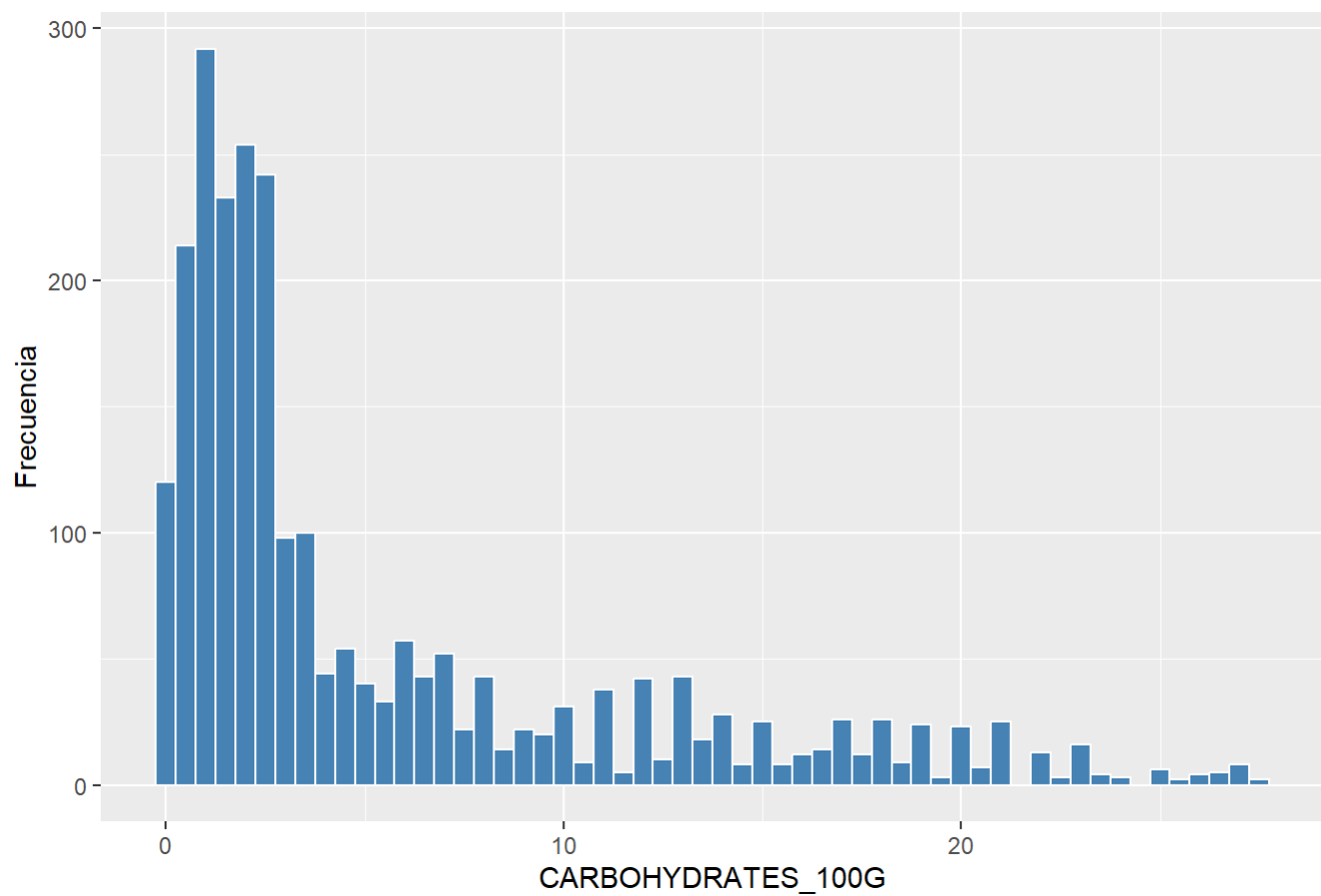
# Crear un histograma para cada columna cuantitativa
for (columna in names(df_tofu[columnas_cuantitativas])) {
  plot_data <- df_tofu[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

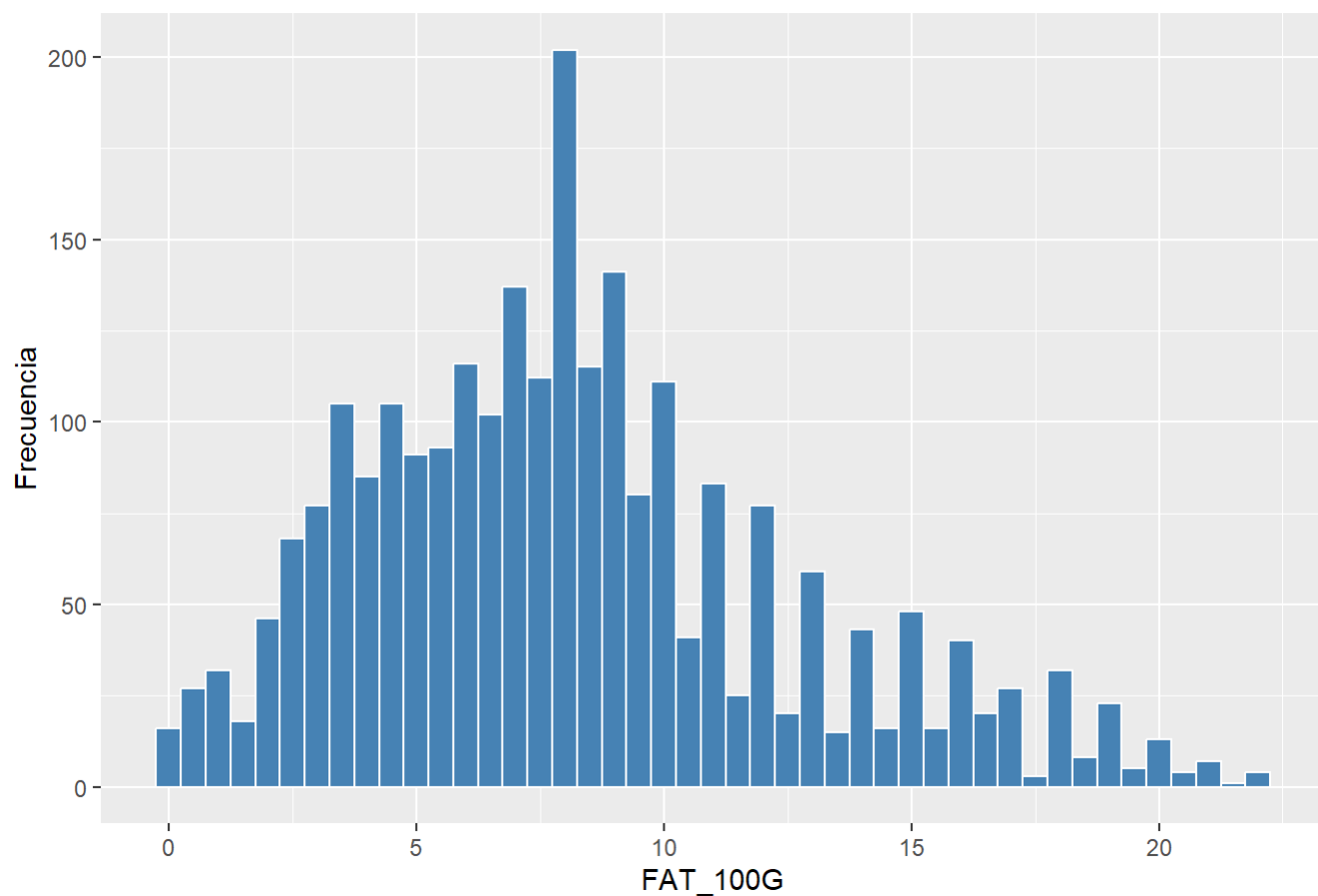
Histograma de PROTEINS_100G



Histograma de CARBOHYDRATES_100G



Histograma de FAT_100G



```
s_tofu <- scale(df_tofu[, -1])
s_tofu1 <- as.data.frame(s_tofu)
summary(s_tofu)
```

```
## PROTEINS_100G CARBOHYDRATES_100G FAT_100G
## Min. :-2.2874 Min. :-0.8833 Min. :-1.88626
## 1st Qu.: -0.7870 1st Qu.: -0.6712 1st Qu.: -0.73328
## Median : 0.1660 Median : -0.4918 Median : -0.08761
## Mean : 0.0000 Mean : 0.0000 Mean : 0.00000
## 3rd Qu.: 0.6932 3rd Qu.: 0.3402 3rd Qu.: 0.48888
## Max. : 2.3964 Max. : 3.5914 Max. : 3.14074
```

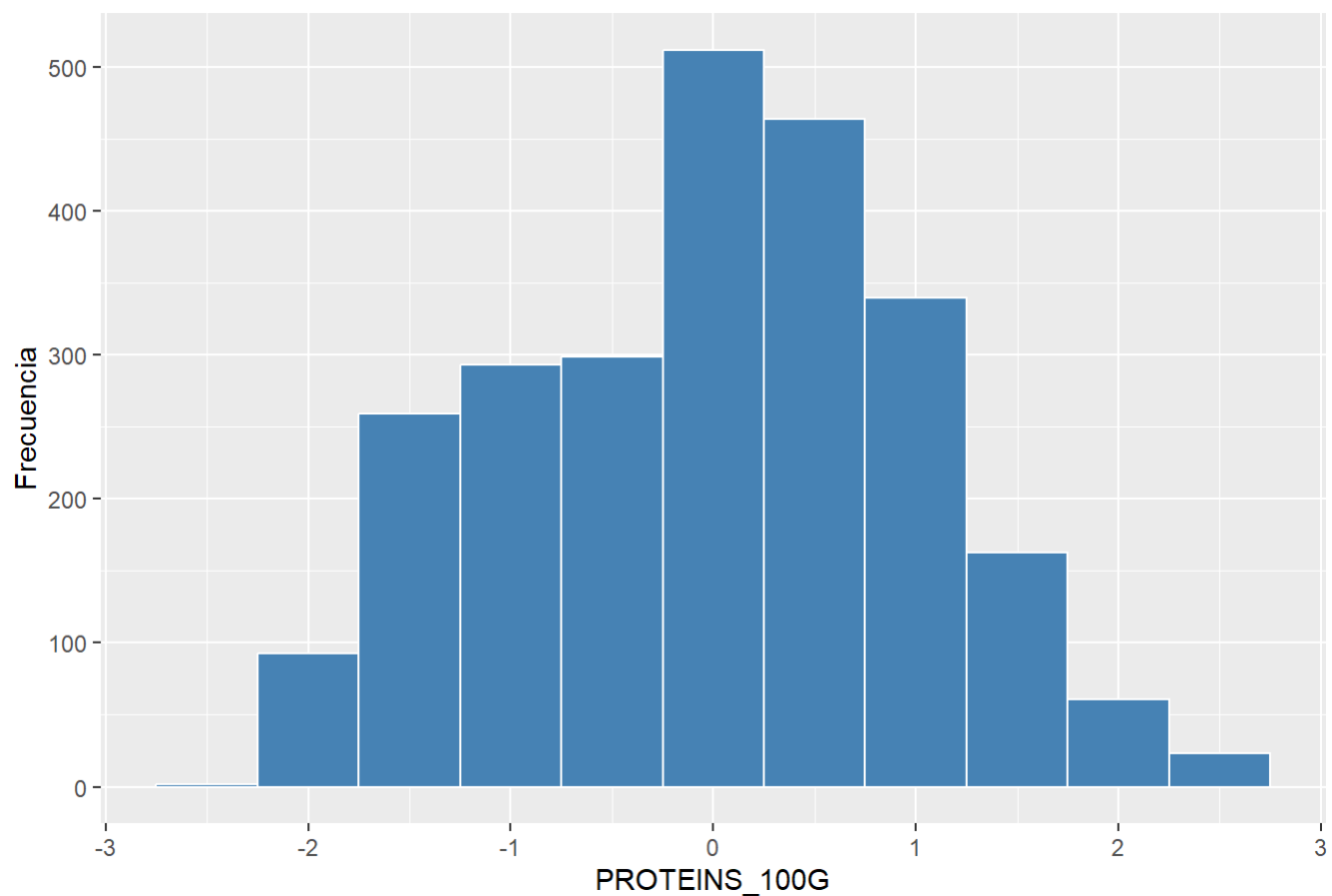
#Datos normalizados para el Tofu

```
# Obtener las columnas cuantitativas del dataframe
columnas_cuantitativas <- sapply(s_tofu1, is.numeric)

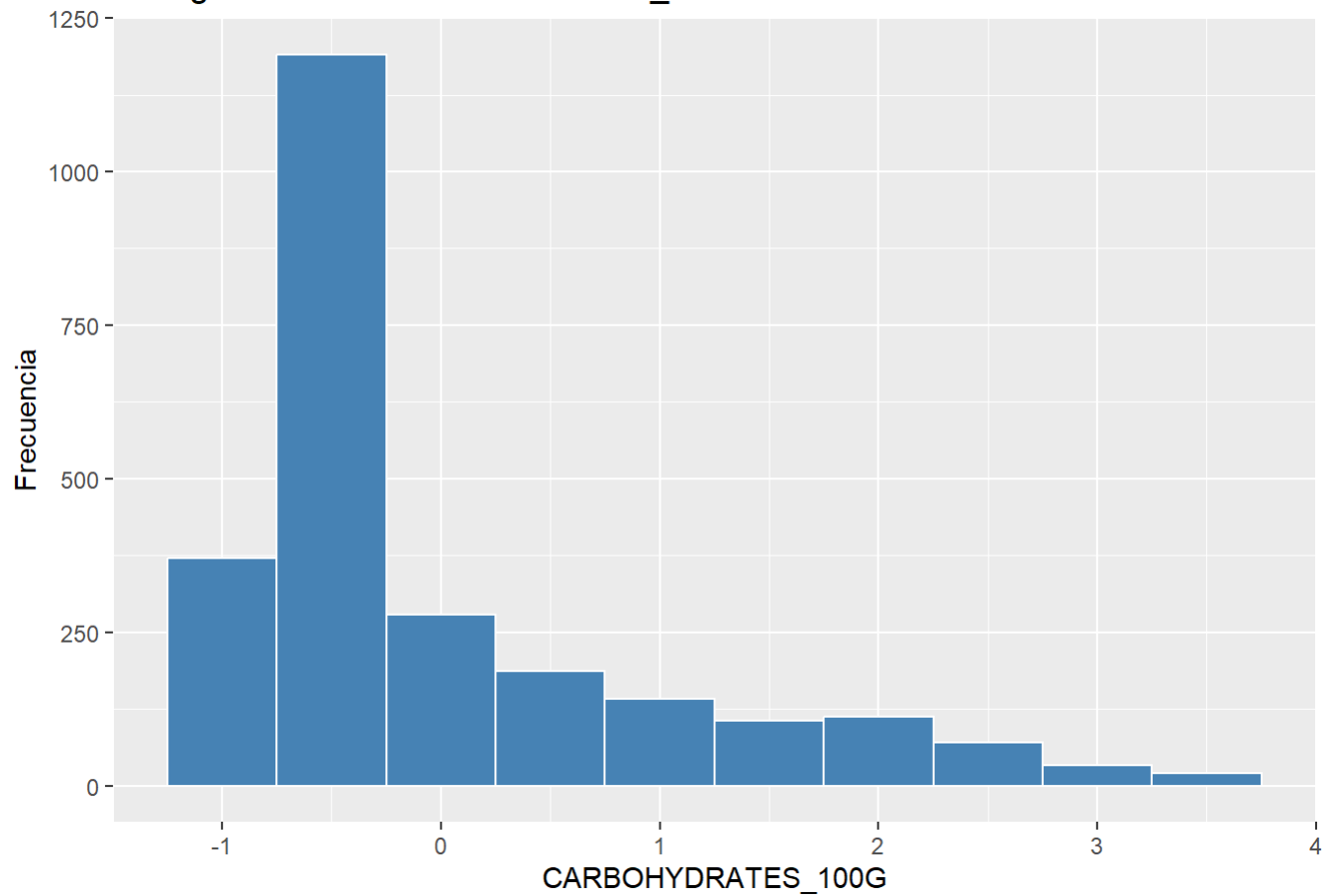
# Crear un histograma para cada columna cuantitativa
for (columna in names(s_tofu1[columnas_cuantitativas])) {
  plot_data <- s_tofu1[, columna]
  p <- ggplot(data.frame(x = plot_data), aes(x)) +
    geom_histogram(binwidth = 0.5, fill = "steelblue", color = "white") +
    labs(title = paste("Histograma de", columna),
         x = columna,
         y = "Frecuencia")

  print(p)
}
```

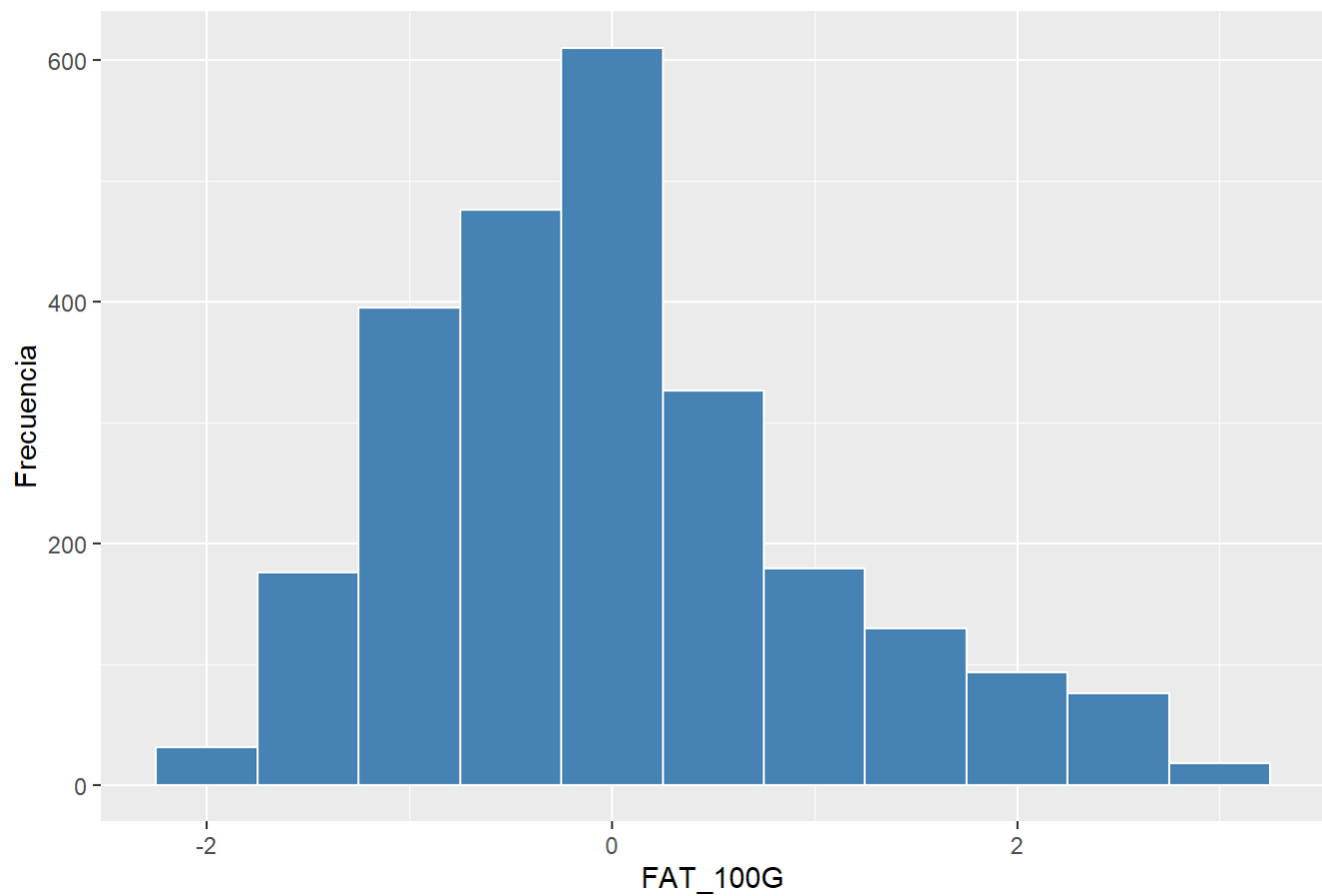
Histograma de PROTEINS_100G



Histograma de CARBOHYDRATES_100G

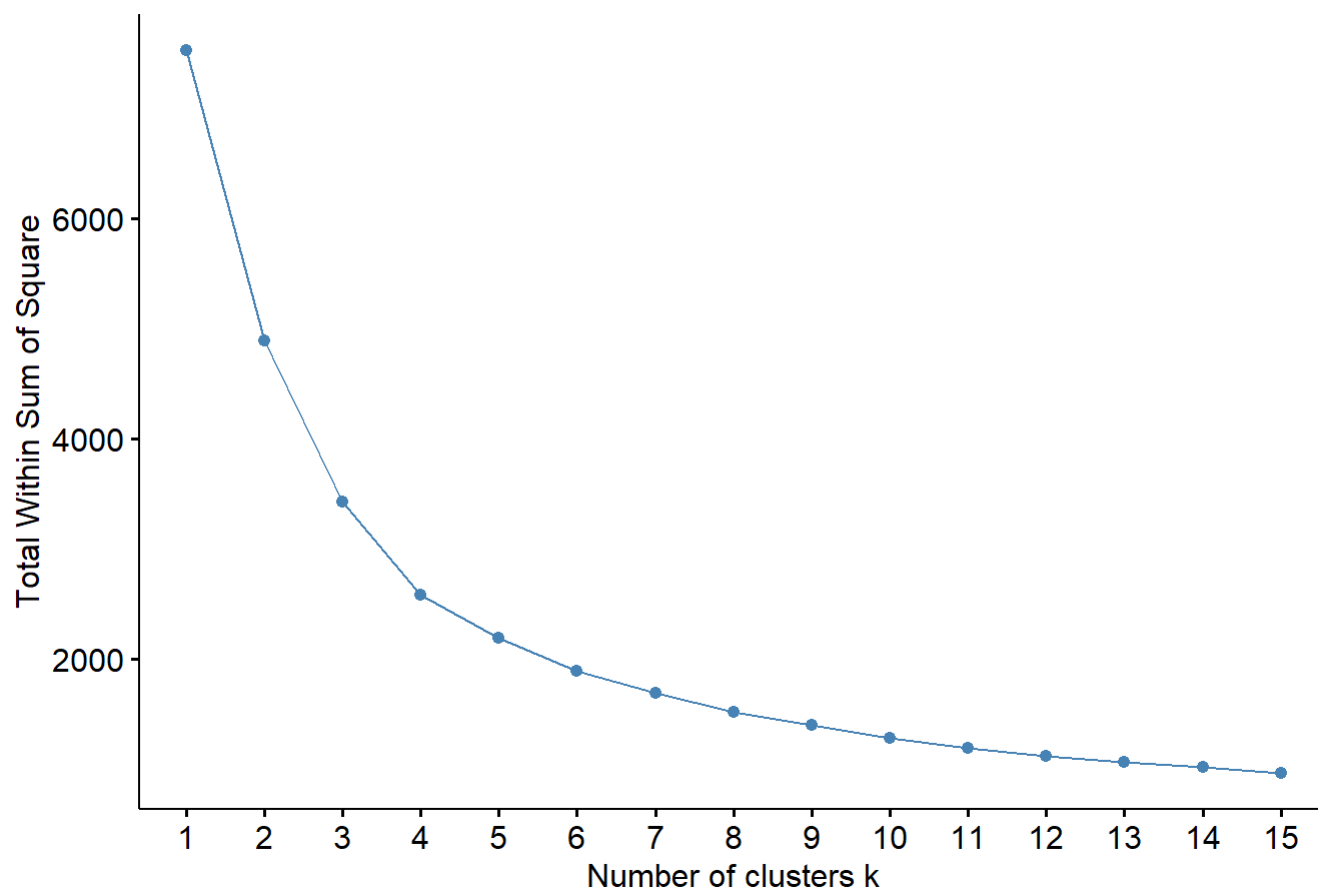


Histograma de FAT_100G



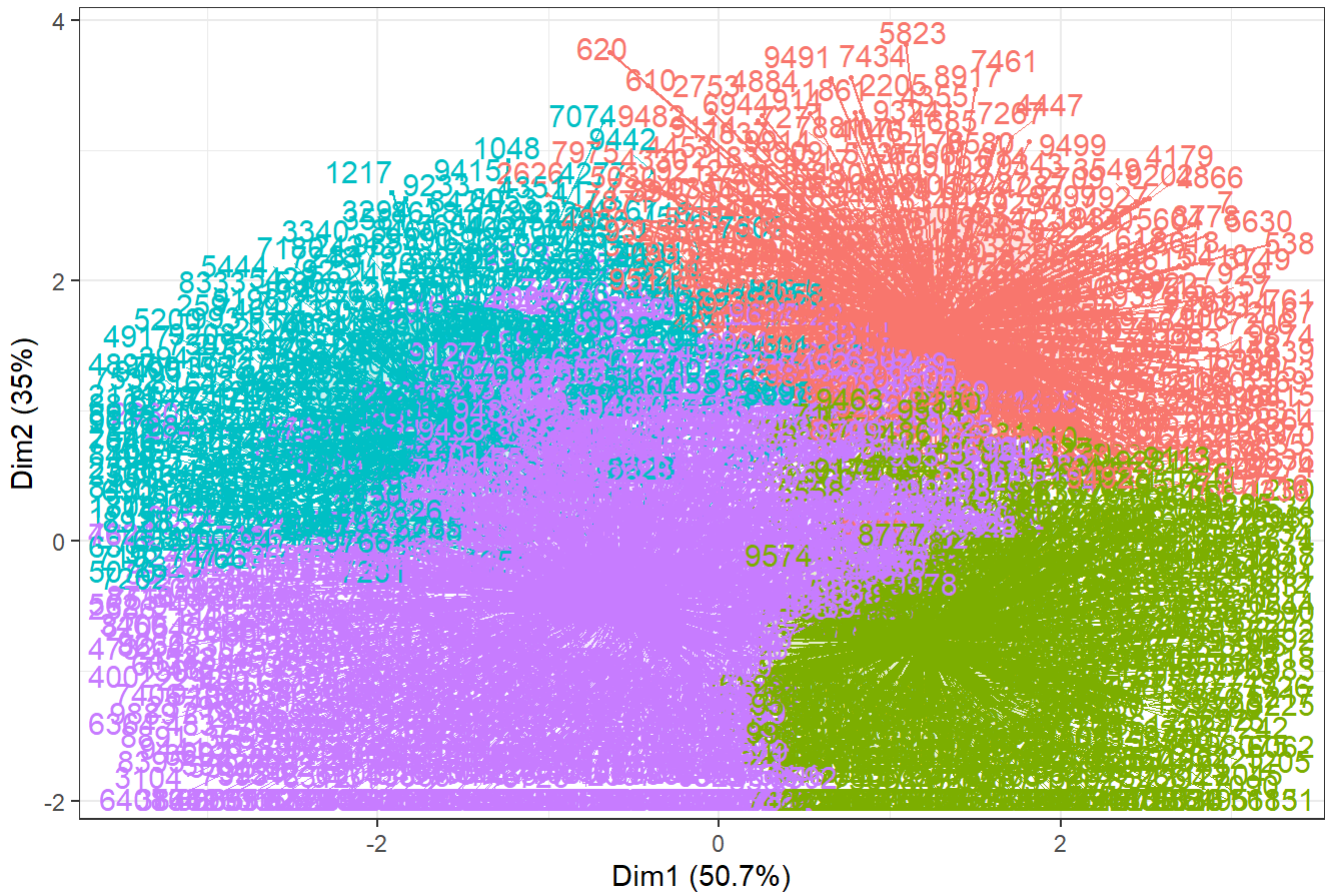
```
s_tofu <- scale(df_tofu[,-1])  
  
# total de cluster óptimos  
elbow <- fviz_nbclust(x = s_tofu, FUNcluster = kmeans, method = "wss", k.max = 15,  
                      diss = get_dist(s_tofu, method = "euclidean"), nstart = 25)  
print(elbow)
```


Optimal number of clusters



```
set.seed(123)
km_clusters <- kmeans(x=s_tofu,centers=4,nstart=25)
fviz_cluster(object=km_clusters,data=s_tofu,show.clust.cent = TRUE,
              ellipse.type="euclid",star.plot=TRUE,repel=TRUE,
              pointsize=0.5,outlier.color="darkred") +
  labs(title = "Resultados clustering K-means") +
  theme_bw() +
  theme(legend.position = "none")
```

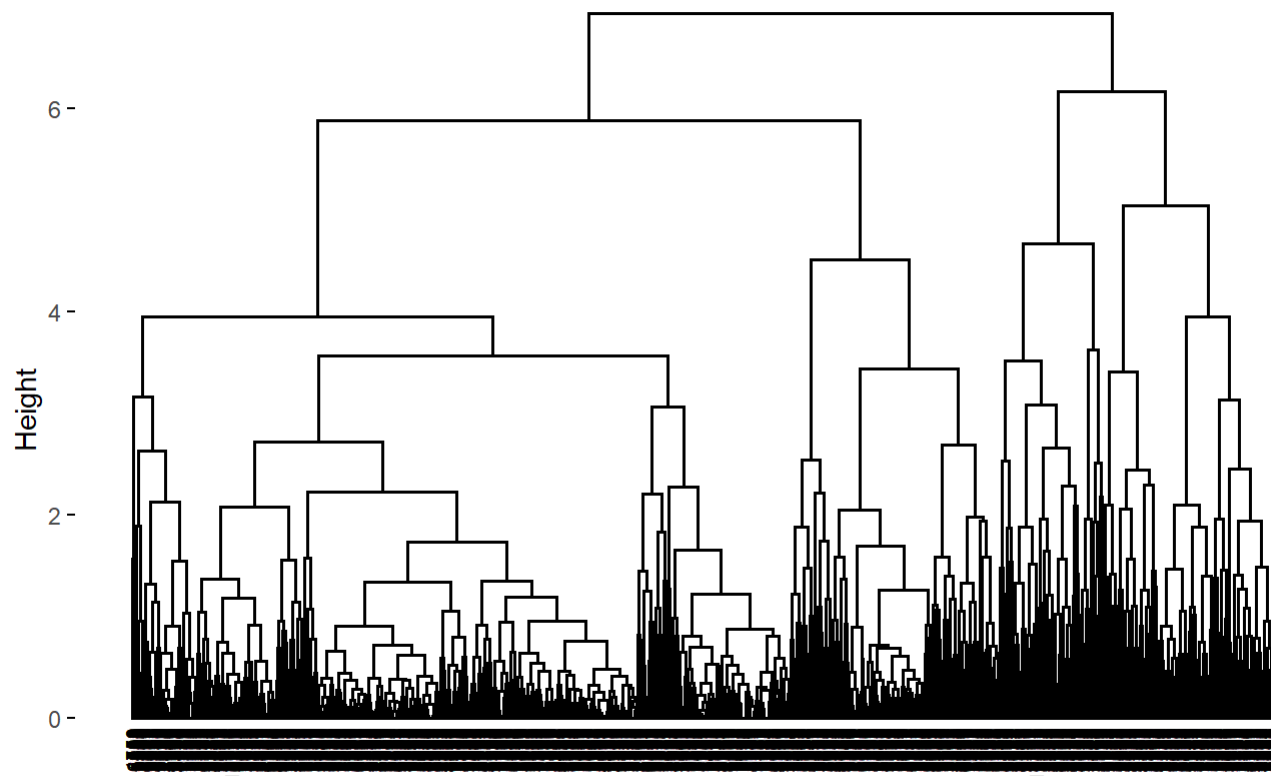
Resultados clustering K-means



```
set.seed(101)
heculidea <- hclust(d = dist(x = s_tofu, method = "euclidean"),
                    method = "complete")

fviz_dend(x=heculidea,cex=0.5,main = "Linkage completo",
          sub="Distancia euclídea")+
  theme(plot.title = element_text(hjust=0.5,size=15))
```

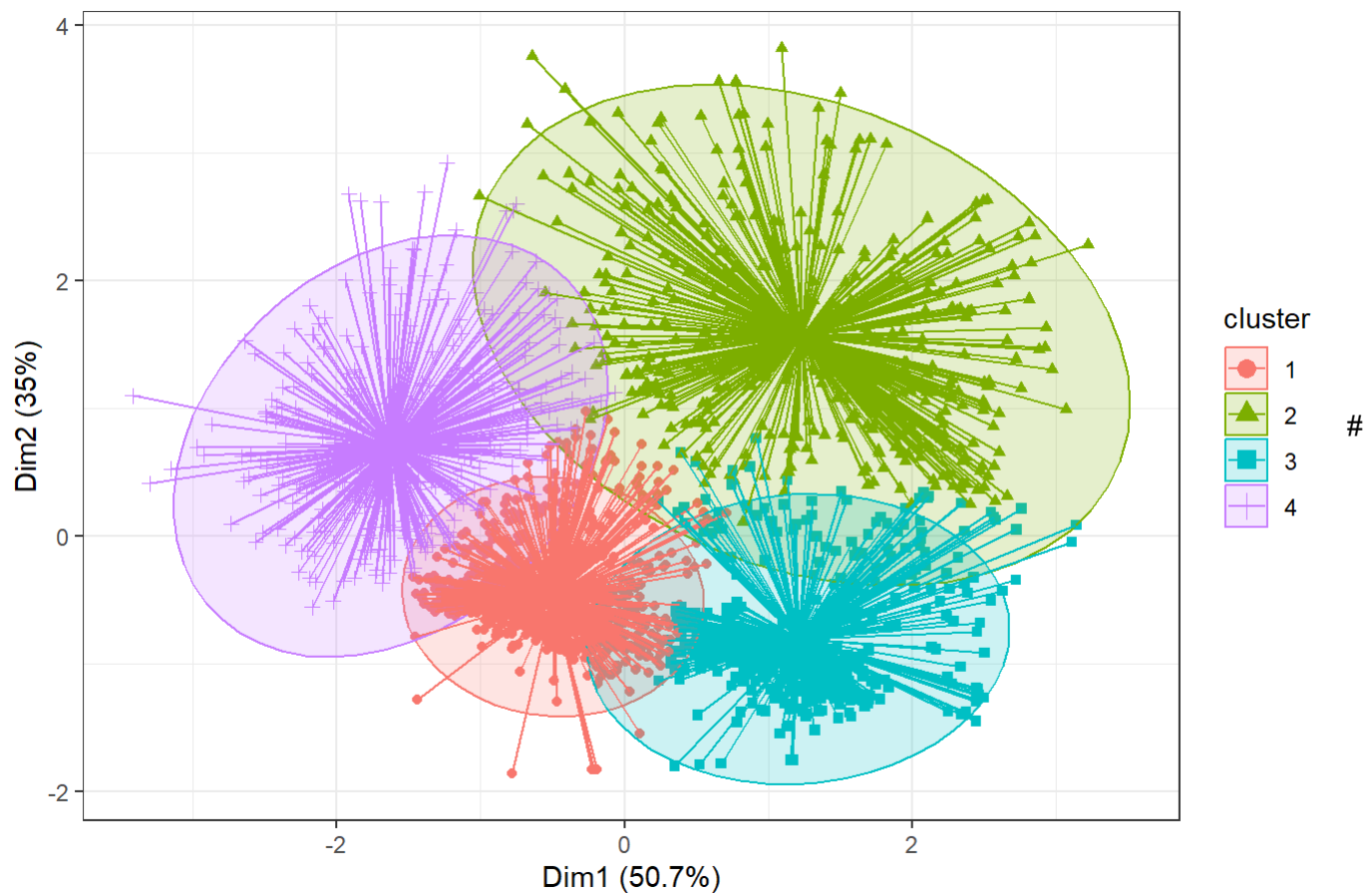
Linkage completo



```
require(cluster)

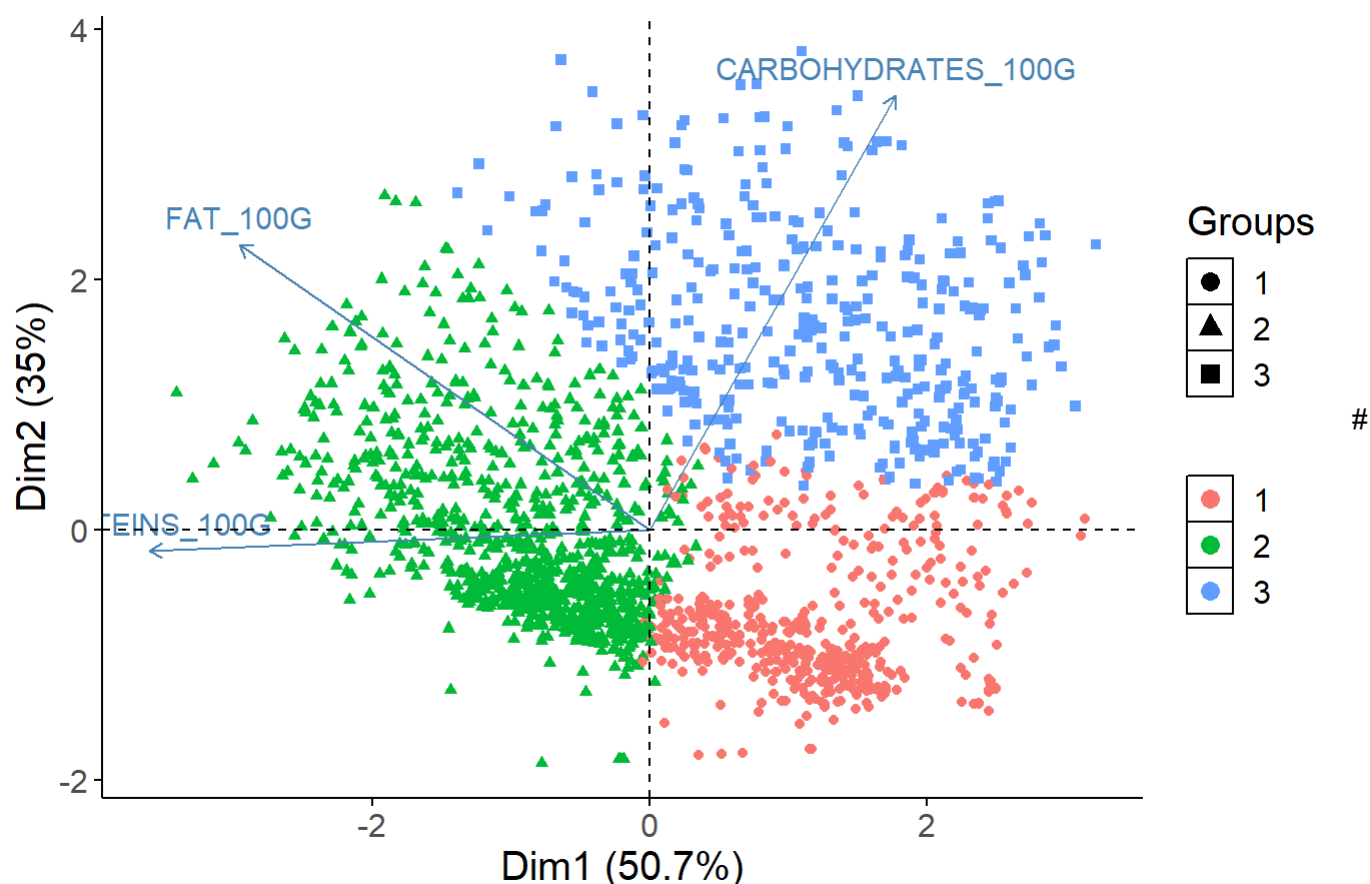
pam.res <- pam(s_tofu, 4)
# Visualización
fviz_cluster(pam.res, geom = "point", ellipse.type = "norm",
              show.clust.cent = TRUE, star.plot = TRUE)+
  labs(title = "Resultados clustering K-means")+ theme_bw()
```

Resultados clustering K-means



Biplot PCA y K-Means para medir representatividad tofu

```
# PCA
pca <- prcomp(df_tofu[, -1], scale=TRUE)
df_tofu.pca <- pca$x
# Cluster over the three first PCA dimensions
kc <- kmeans(df_tofu.pca[, 1:3], 3)
fviz_pca_biplot(pca, label="var", habillage=as.factor(kc$cluster)) +
  labs(color=NULL) + ggtitle("") +
  theme(text = element_text(size = 15),
        panel.background = element_blank(),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        axis.line = element_line(colour = "black"),
        legend.key = element_rect(fill = "white"))
```



Método no jerárquico CLARA basado en simulación y muestreo

Soja

```
clara_clusters <- clara(x = scale(df_soja[,-1]), k = 4, metric = "manhattan", stand = TRUE,
                        samples = 60, pamLike = TRUE)
clara_clusters$sample
```

```
## [1] "1013" "1032" "1345" "1576" "1597" "1737" "1796" "1830" "1884"
## [10] "2161" "2345" "2446" "2490" "2513" "2911" "3002" "3544" "3598"
## [19] "3635" "3991" "4055" "4178" "4223" "4297" "4299" "4402" "4698"
## [28] "5522" "5835" "6028" "6136" "6174" "6424" "6493" "6578" "6884"
## [37] "6978" "7006" "7525" "7552" "7633" "7672" "8437" "8631" "8671"
## [46] "9004" "9397" "10003"
```

```
clara_clusters$medoids
```

```
##      PROTEINS_100G CARBOHYDRATES_100G  FAT_100G
## 8671   -0.1011949         1.8234211  0.76401714
## 4178    0.1431987        -0.5813796  0.55109662
## 1345   -0.6782355        -0.3889956 -0.69803713
## 7552    2.3834738         0.3805407  0.06847676
```

```
clara_clusters$i.med
```

```
## [1] 1991  963  248 1771
```

```
clara_clusters$clustering
```

##	11	23	24	26	28	37	39	56	90	188	223	253	294
##	1	1	2	2	3	3	2	3	3	1	3	3	3
##	295	298	308	317	332	339	345	348	352	364	365	372	374
##	1	4	3	1	2	2	3	3	3	3	4	2	2
##	375	377	379	380	384	386	389	401	404	405	408	411	419
##	3	1	3	2	3	3	3	3	4	2	2	3	3
##	422	433	440	449	450	476	479	480	485	507	540	552	570
##	3	4	3	1	3	3	3	3	4	3	4	3	3
##	574	585	647	659	663	667	669	677	681	683	704	709	714
##	2	1	1	2	4	2	3	3	3	4	2	1	3
##	728	729	732	737	743	744	751	762	763	767	771	778	779
##	2	2	2	2	2	2	3	2	3	3	2	4	3
##	780	783	786	790	796	799	803	805	823	825	833	843	847
##	2	3	2	3	2	4	3	3	1	3	3	3	3
##	849	851	854	873	880	882	892	897	932	935	944	959	962
##	2	2	1	3	3	3	2	2	4	3	3	3	2
##	965	966	967	968	978	987	991	992	994	995	999	1000	1001
##	2	3	3	2	3	2	3	2	3	2	3	2	1
##	1008	1009	1010	1012	1013	1014	1018	1019	1023	1024	1025	1026	1029
##	1	2	2	2	1	3	3	3	2	2	2	2	2
##	1030	1032	1035	1037	1040	1041	1042	1047	1049	1054	1057	1058	1060
##	3	3	3	3	3	3	3	2	2	3	3	1	2
##	1062	1064	1066	1068	1072	1073	1074	1078	1079	1080	1081	1086	1089
##	2	2	3	1	3	2	2	2	3	3	2	3	2
##	1091	1092	1094	1095	1101	1103	1105	1110	1116	1117	1118	1122	1124
##	2	3	1	2	2	3	3	2	3	1	3	3	3
##	1129	1130	1131	1132	1134	1135	1136	1140	1141	1146	1147	1148	1153
##	3	2	1	1	2	3	2	3	2	2	1	3	3
##	1160	1165	1169	1172	1173	1175	1176	1179	1181	1184	1185	1186	1187
##	2	2	3	3	3	2	2	2	1	1	2	3	3
##	1188	1189	1192	1193	1197	1200	1208	1212	1214	1216	1219	1220	1227
##	2	2	2	3	3	3	3	3	3	3	2	3	3
##	1232	1233	1234	1238	1243	1245	1249	1250	1251	1252	1254	1256	1261
##	4	3	2	3	2	3	3	3	1	2	3	3	2
##	1264	1267	1268	1271	1272	1276	1281	1284	1285	1286	1294	1302	1307
##	2	2	2	2	1	3	1	3	3	3	3	3	3
##	1312	1315	1316	1318	1327	1328	1331	1333	1337	1338	1339	1342	1344
##	2	3	2	2	3	3	3	3	2	4	3	3	2
##	1345	1346	1348	1350	1354	1355	1356	1357	1364	1366	1367	1369	1374
##	3	2	3	2	3	3	3	3	3	3	3	3	3
##	1375	1376	1377	1378	1379	1381	1384	1385	1388	1390	1392	1393	1395
##	3	3	3	3	3	2	3	2	2	3	1	3	2
##	1396	1398	1400	1401	1402	1403	1404	1405	1406	1409	1410	1411	1414
##	3	2	3	3	3	3	3	4	3	3	2	3	3
##	1415	1417	1418	1419	1422	1423	1430	1432	1433	1435	1436	1438	1440
##	3	3	3	3	1	3	2	3	3	3	3	3	3
##	1441	1442	1444	1449	1452	1454	1455	1456	1458	1459	1460	1462	1464
##	2	3	3	3	3	1	2	3	2	3	3	3	3
##	1465	1467	1470	1471	1472	1473	1474	1476	1477	1481	1482	1484	1486
##	3	3	3	3	3	3	2	3	3	3	2	2	3
##	1487	1489	1490	1492	1494	1495	1497	1498	1500	1502	1504	1505	1507
##	3	3	3	3	2	2	3	3	2	1	3	3	1

##	1508	1510	1515	1516	1517	1523	1525	1526	1527	1528	1529	1530	1533
##	3	2	4	3	2	3	3	3	1	3	3	3	1
##	1535	1536	1537	1540	1543	1547	1548	1549	1551	1557	1558	1559	1560
##	1	2	3	3	1	3	3	4	3	3	3	3	2
##	1562	1564	1566	1567	1568	1570	1575	1576	1585	1588	1592	1593	1596
##	1	4	3	4	1	4	3	2	2	2	4	4	4
##	1597	1601	1608	1609	1620	1623	1624	1626	1629	1630	1631	1635	1640
##	3	4	2	3	4	2	4	2	2	4	1	3	3
##	1642	1645	1649	1650	1656	1659	1662	1666	1667	1669	1674	1677	1679
##	3	3	3	1	3	2	3	1	3	1	3	3	3
##	1680	1683	1684	1685	1688	1690	1691	1696	1698	1699	1700	1701	1708
##	2	2	2	2	2	3	2	3	1	2	3	3	3
##	1709	1711	1712	1713	1714	1721	1722	1723	1725	1726	1727	1728	1729
##	2	2	3	3	1	2	2	3	3	3	3	2	2
##	1734	1737	1741	1743	1746	1749	1751	1752	1758	1760	1761	1763	1778
##	3	1	2	3	1	3	1	3	3	2	3	2	2
##	1781	1782	1790	1793	1794	1796	1797	1801	1805	1810	1812	1814	1815
##	2	1	3	2	3	2	2	2	3	3	3	1	2
##	1816	1817	1818	1821	1823	1827	1828	1830	1831	1833	1836	1839	1845
##	3	3	1	2	3	3	3	3	2	3	3	3	2
##	1849	1853	1855	1857	1869	1873	1875	1876	1877	1878	1884	1885	1889
##	3	1	2	4	4	3	1	3	3	2	3	3	2
##	1892	1893	1898	1904	1907	1911	1916	1927	1928	1931	1940	1945	1948
##	3	3	2	3	1	3	3	1	1	2	1	1	1
##	1953	1962	1965	1966	1978	1988	1989	1990	1998	2003	2006	2009	2010
##	1	4	1	1	2	3	2	4	4	4	4	4	4
##	2011	2015	2019	2022	2024	2026	2033	2035	2037	2040	2041	2047	2048
##	2	4	4	1	4	1	4	1	4	1	4	4	4
##	2053	2055	2057	2058	2061	2062	2070	2075	2077	2085	2088	2093	2097
##	4	2	3	4	3	4	3	1	2	3	4	3	3
##	2099	2100	2103	2105	2106	2109	2119	2131	2133	2141	2143	2158	2161
##	4	3	4	2	2	3	2	2	4	3	3	3	2
##	2162	2165	2175	2177	2181	2182	2184	2193	2203	2204	2211	2219	2220
##	4	3	3	1	3	1	4	2	4	2	4	3	2
##	2223	2239	2242	2252	2261	2266	2268	2269	2271	2273	2274	2275	2277
##	2	3	3	3	3	2	4	2	3	3	4	2	3
##	2279	2288	2292	2294	2299	2300	2307	2312	2316	2320	2324	2325	2326
##	3	2	3	1	2	3	2	3	3	3	3	4	3
##	2328	2334	2335	2338	2344	2345	2350	2354	2360	2361	2363	2366	2370
##	3	1	2	4	3	4	4	2	2	4	2	4	3
##	2371	2373	2374	2384	2385	2386	2391	2392	2396	2398	2400	2409	2413
##	3	3	4	4	3	4	2	4	4	4	3	3	4
##	2422	2427	2429	2432	2437	2446	2452	2455	2457	2460	2462	2464	2465
##	4	3	4	1	2	2	4	1	1	3	2	3	4
##	2466	2467	2469	2473	2475	2476	2477	2483	2484	2488	2490	2491	2492
##	2	4	4	3	2	3	2	3	2	3	4	3	3
##	2495	2496	2497	2498	2502	2503	2509	2513	2520	2523	2525	2526	2529
##	4	4	4	4	2	3	2	2	4	4	3	4	4
##	2534	2536	2537	2539	2544	2548	2550	2555	2556	2558	2563	2564	2567
##	3	4	2	3	2	3	2	3	4	2	2	3	2
##	2574	2579	2580	2581	2585	2593	2597	2603	2604	2605	2609	2611	2618
##	3	4	4	3	4	1	4	4	2	2	4	2	2

##	2622	2625	2633	2635	2638	2665	2667	2677	2699	2733	2734	2740	2762
##	2	2	1	3	1	3	3	3	2	1	3	3	1
##	2775	2781	2792	2813	2838	2842	2850	2852	2877	2879	2880	2883	2893
##	4	3	2	4	1	3	2	3	3	3	3	1	3
##	2898	2902	2905	2911	2912	2914	2915	2916	2922	2932	2934	2935	2936
##	2	3	3	2	3	3	1	3	3	3	3	3	3
##	2951	2955	2958	2967	2971	2972	2977	2978	2985	2992	2996	3002	3003
##	3	3	3	2	3	3	3	2	3	2	2	1	3
##	3008	3011	3014	3016	3023	3026	3029	3033	3034	3037	3039	3041	3047
##	3	3	3	2	3	2	3	3	1	3	3	2	3
##	3054	3055	3056	3067	3069	3078	3080	3091	3092	3098	3100	3107	3108
##	3	3	2	3	3	2	3	2	2	3	2	1	2
##	3116	3119	3123	3126	3127	3134	3136	3139	3142	3143	3144	3146	3147
##	3	3	2	2	4	2	2	3	3	4	3	4	1
##	3149	3156	3160	3163	3165	3168	3178	3180	3181	3186	3187	3188	3190
##	4	1	4	3	3	4	3	1	4	1	2	2	4
##	3192	3196	3199	3200	3201	3215	3231	3239	3254	3260	3263	3265	3269
##	3	3	3	4	4	2	2	3	2	2	2	2	3
##	3276	3288	3292	3298	3299	3315	3321	3322	3323	3333	3334	3348	3356
##	2	3	2	4	3	3	3	3	2	1	2	1	3
##	3362	3379	3380	3395	3403	3406	3412	3419	3428	3433	3439	3445	3462
##	1	2	3	2	2	3	2	2	3	2	3	4	3
##	3465	3486	3491	3492	3493	3494	3498	3503	3504	3511	3516	3529	3534
##	3	2	4	2	2	3	2	2	2	3	3	3	3
##	3543	3544	3550	3553	3560	3562	3565	3567	3569	3571	3572	3573	3574
##	2	2	2	2	3	2	2	2	2	2	2	2	3
##	3576	3577	3579	3587	3588	3596	3598	3600	3605	3606	3608	3609	3610
##	2	1	2	3	3	3	2	3	1	3	2	2	2
##	3611	3613	3614	3619	3622	3623	3626	3628	3635	3640	3650	3659	3660
##	3	3	3	2	3	3	4	3	3	4	3	3	1
##	3664	3665	3670	3675	3680	3681	3689	3697	3705	3706	3707	3708	3714
##	2	4	3	3	2	3	1	3	3	3	4	2	2
##	3715	3718	3720	3728	3730	3731	3732	3736	3740	3742	3751	3756	3757
##	3	3	2	1	4	1	3	1	4	1	3	3	3
##	3764	3767	3769	3774	3775	3781	3791	3818	3820	3839	3874	3903	3911
##	3	3	3	3	4	1	1	3	2	1	3	1	2
##	3925	3927	3930	3934	3946	3953	3957	3960	3964	3969	3971	3975	3989
##	3	3	4	3	3	3	3	3	3	3	3	4	3
##	3990	3991	3992	4000	4007	4013	4014	4018	4020	4024	4030	4032	4033
##	3	2	3	3	3	1	3	2	1	3	1	3	3
##	4035	4039	4042	4044	4045	4051	4055	4062	4067	4069	4070	4071	4074
##	4	3	2	3	3	3	3	1	1	3	4	3	1
##	4078	4084	4096	4114	4132	4135	4141	4143	4145	4147	4149	4156	4170
##	1	4	3	1	1	3	4	2	2	2	4	1	3
##	4178	4180	4184	4187	4189	4196	4203	4206	4209	4210	4219	4220	4223
##	2	2	4	1	2	2	4	3	3	4	4	3	3
##	4233	4240	4246	4257	4271	4275	4278	4282	4288	4293	4297	4299	4307
##	3	2	2	3	4	4	3	4	3	1	2	3	4
##	4310	4317	4328	4329	4338	4344	4346	4350	4361	4363	4374	4375	4391
##	1	2	3	3	3	4	3	3	1	3	3	3	3
##	4402	4413	4414	4417	4425	4426	4430	4431	4444	4454	4458	4463	4469
##	4	1	2	2	4	3	2	3	3	1	4	4	1

##	4471	4473	4493	4498	4500	4516	4529	4531	4533	4538	4557	4562	4564
##	3	3	4	2	4	3	4	4	1	1	4	3	1
##	4567	4569	4573	4577	4579	4584	4586	4588	4592	4598	4603	4606	4612
##	2	3	1	3	3	2	2	3	3	1	3	2	3
##	4623	4628	4638	4652	4656	4657	4680	4686	4687	4688	4689	4698	4706
##	3	3	3	3	3	4	3	2	2	2	3	3	1
##	4707	4714	4715	4716	4727	4742	4751	4768	4774	4776	4784	4787	4792
##	3	2	3	3	3	3	2	4	2	3	2	2	4
##	4805	4814	4835	4841	4851	4855	4856	4861	4864	4868	4875	4877	4878
##	3	2	3	2	3	3	2	2	4	3	3	4	3
##	4894	4895	4902	4903	4907	4909	4912	4915	4918	4921	4928	4942	4962
##	4	2	2	2	3	3	3	4	4	1	3	2	2
##	4969	4972	4982	4984	4990	5038	5042	5075	5093	5109	5111	5120	5128
##	2	2	2	1	3	4	4	3	3	1	4	3	3
##	5131	5226	5249	5261	5273	5281	5294	5315	5316	5317	5318	5321	5325
##	3	1	3	3	1	1	3	3	3	3	1	4	2
##	5327	5338	5340	5342	5345	5346	5348	5351	5352	5353	5358	5359	5362
##	3	4	2	3	2	3	2	3	3	2	2	1	2
##	5375	5380	5392	5400	5406	5418	5420	5422	5441	5442	5451	5466	5487
##	4	2	3	2	3	1	2	2	3	4	2	2	3
##	5491	5498	5505	5507	5522	5534	5535	5559	5603	5650	5654	5660	5678
##	1	3	3	3	2	4	3	3	3	3	3	2	3
##	5679	5680	5682	5684	5689	5699	5705	5707	5709	5710	5720	5729	5730
##	3	2	3	1	2	3	3	3	3	3	2	4	2
##	5732	5733	5734	5735	5737	5740	5756	5757	5762	5775	5779	5784	5793
##	4	2	2	3	2	1	3	2	3	2	3	3	2
##	5796	5797	5799	5801	5804	5806	5808	5820	5822	5835	5836	5839	5840
##	3	2	3	3	3	3	3	3	1	3	3	3	2
##	5841	5851	5853	5856	5861	5864	5868	5878	5885	5900	5929	5931	5940
##	3	3	3	3	3	1	4	3	4	3	2	1	2
##	5950	5952	5962	5963	5973	5977	5978	5981	5984	5985	5987	5991	5992
##	1	3	2	3	2	1	2	3	3	2	1	3	3
##	5994	5995	6003	6004	6006	6007	6008	6010	6013	6019	6020	6023	6028
##	3	1	2	2	3	3	3	3	2	2	2	2	3
##	6029	6032	6033	6034	6037	6039	6040	6041	6042	6043	6045	6047	6049
##	2	2	3	2	3	3	3	4	2	2	3	3	3
##	6051	6058	6060	6061	6064	6065	6068	6069	6072	6074	6076	6078	6080
##	3	2	3	3	2	2	3	2	2	3	3	1	3
##	6084	6085	6087	6088	6090	6096	6098	6099	6100	6101	6102	6103	6104
##	2	1	2	3	3	1	2	3	2	2	3	2	1
##	6105	6107	6108	6112	6114	6115	6118	6120	6123	6127	6130	6131	6132
##	2	2	2	3	3	2	2	3	2	3	3	3	3
##	6133	6136	6137	6138	6140	6141	6147	6148	6150	6152	6153	6154	6157
##	3	3	2	3	3	3	3	2	2	1	2	4	2
##	6161	6163	6164	6171	6174	6175	6176	6177	6179	6183	6185	6186	6188
##	2	1	2	2	3	3	3	2	3	2	4	2	3
##	6189	6192	6200	6204	6207	6210	6211	6212	6217	6218	6219	6221	6223
##	2	3	2	3	2	3	1	2	3	3	2	3	3
##	6226	6227	6231	6232	6233	6236	6238	6239	6240	6242	6245	6249	6250
##	2	3	3	3	3	1	2	3	3	3	4	3	3
##	6251	6253	6254	6256	6258	6262	6263	6266	6271	6272	6274	6276	6278
##	3	2	3	3	3	2	1	3	2	3	3	3	3

##	6284	6285	6286	6287	6289	6292	6294	6303	6305	6306	6315	6323	6324
##	3	3	3	3	3	3	3	1	4	2	3	3	2
##	6328	6329	6330	6331	6332	6334	6336	6337	6339	6344	6346	6347	6350
##	1	3	3	4	1	3	2	3	3	2	2	3	1
##	6352	6355	6360	6362	6365	6370	6371	6377	6379	6380	6383	6386	6387
##	3	2	3	3	2	3	3	3	3	2	3	3	3
##	6388	6389	6391	6397	6399	6403	6404	6406	6407	6408	6409	6411	6416
##	3	3	3	3	3	3	3	2	3	2	2	3	3
##	6417	6418	6419	6420	6421	6424	6428	6430	6431	6434	6437	6438	6439
##	4	2	3	3	3	3	2	2	2	1	2	3	3
##	6440	6442	6443	6445	6446	6447	6448	6452	6453	6455	6458	6459	6464
##	2	3	3	3	2	2	2	3	3	3	1	2	3
##	6467	6469	6471	6474	6476	6477	6478	6482	6483	6484	6487	6488	6489
##	3	3	3	3	3	3	3	2	3	3	3	3	3
##	6491	6492	6493	6495	6496	6497	6498	6499	6500	6501	6504	6505	6507
##	3	3	3	2	3	3	3	3	2	3	3	3	4
##	6509	6512	6516	6517	6518	6521	6523	6525	6526	6527	6529	6531	6538
##	2	3	3	2	2	2	3	3	3	3	2	3	3
##	6539	6540	6541	6543	6544	6545	6549	6550	6551	6552	6553	6554	6555
##	3	2	3	2	2	2	3	3	3	3	3	3	3
##	6556	6557	6559	6560	6563	6564	6567	6574	6575	6576	6578	6583	6585
##	3	3	3	2	3	1	3	3	3	3	4	4	2
##	6586	6589	6590	6594	6598	6603	6607	6611	6614	6616	6620	6622	6624
##	3	4	4	2	2	3	3	2	2	3	2	1	3
##	6625	6627	6630	6632	6633	6639	6640	6649	6650	6652	6653	6656	6658
##	2	4	3	3	4	2	3	3	4	3	3	4	3
##	6664	6667	6668	6669	6672	6674	6677	6679	6681	6682	6685	6686	6693
##	2	4	3	4	2	3	3	3	2	2	1	4	3
##	6696	6698	6704	6705	6707	6710	6712	6715	6719	6730	6734	6736	6740
##	3	4	4	4	3	3	3	2	3	2	3	4	2
##	6750	6751	6752	6753	6754	6756	6758	6759	6760	6762	6763	6767	6770
##	2	3	3	2	3	2	2	1	3	1	2	1	2
##	6776	6777	6778	6784	6788	6789	6790	6792	6794	6795	6797	6801	6802
##	2	3	2	2	3	3	2	3	3	2	2	3	3
##	6805	6806	6809	6810	6811	6812	6813	6814	6820	6821	6824	6828	6832
##	3	3	3	3	3	2	3	3	3	2	3	3	3
##	6834	6835	6839	6841	6845	6847	6848	6849	6853	6857	6859	6860	6861
##	2	2	3	3	3	1	1	3	3	3	3	2	1
##	6862	6863	6864	6867	6870	6875	6876	6877	6879	6884	6885	6886	6889
##	2	2	2	2	3	3	2	3	3	3	2	2	3
##	6890	6892	6894	6896	6903	6904	6905	6910	6911	6912	6913	6920	6923
##	3	3	3	2	2	2	3	3	2	3	2	3	3
##	6925	6927	6930	6938	6942	6948	6951	6952	6957	6958	6967	6969	6972
##	3	2	3	3	1	2	3	3	3	2	2	2	4
##	6973	6976	6978	6986	6991	6996	7001	7003	7006	7008	7013	7016	7020
##	4	1	3	3	1	3	2	3	2	1	4	3	2
##	7023	7025	7034	7035	7042	7043	7047	7048	7049	7051	7058	7063	7064
##	3	2	4	2	3	4	4	2	4	3	4	3	4
##	7065	7066	7070	7071	7075	7082	7096	7101	7102	7111	7115	7120	7121
##	1	4	4	4	2	4	3	3	4	4	3	4	3
##	7122	7128	7131	7133	7137	7142	7143	7144	7147	7150	7165	7167	7175
##	4	4	2	4	3	4	1	4	4	3	2	1	3

##	7177	7181	7185	7186	7214	7230	7234	7240	7247	7250	7256	7257	7259
##	1	3	2	4	2	2	2	1	3	4	2	3	4
##	7272	7273	7281	7284	7286	7287	7289	7290	7296	7303	7305	7308	7315
##	4	3	2	2	3	2	2	1	2	4	3	3	4
##	7323	7324	7329	7330	7333	7337	7340	7342	7347	7349	7352	7355	7356
##	2	3	2	2	3	3	3	4	3	2	3	3	3
##	7362	7363	7369	7372	7374	7379	7381	7389	7391	7392	7395	7396	7412
##	3	3	3	3	3	3	3	3	3	3	3	3	3
##	7413	7422	7440	7443	7447	7451	7456	7460	7462	7470	7471	7473	7479
##	3	4	4	4	3	2	2	3	4	3	2	3	3
##	7480	7486	7487	7488	7490	7496	7501	7521	7525	7537	7541	7545	7546
##	3	4	3	2	3	4	4	4	4	3	3	4	3
##	7547	7550	7552	7555	7557	7559	7560	7567	7569	7570	7571	7574	7576
##	2	3	4	4	3	4	2	3	4	4	1	3	2
##	7579	7580	7582	7588	7591	7595	7596	7597	7599	7611	7617	7624	7633
##	2	2	2	4	3	4	3	4	2	3	2	2	2
##	7634	7640	7655	7656	7668	7671	7672	7677	7678	7680	7686	7703	7713
##	2	4	1	4	4	2	3	3	4	4	2	1	3
##	7717	7729	7741	7742	7745	7754	7756	7758	7771	7773	7776	7781	7786
##	1	3	1	2	1	2	1	1	1	4	3	1	3
##	7790	7792	7795	7796	7807	7810	7812	7815	7820	7827	7843	7887	7890
##	1	1	1	1	1	2	3	3	2	3	2	3	3
##	7907	7945	7946	7963	7964	7969	7970	7976	7983	7985	7996	7997	8004
##	3	2	1	3	2	3	3	3	3	3	3	2	3
##	8009	8016	8018	8022	8024	8025	8027	8031	8040	8044	8046	8048	8050
##	1	2	1	3	3	3	2	3	3	2	3	3	3
##	8058	8076	8078	8080	8087	8091	8092	8099	8105	8114	8131	8138	8141
##	2	3	3	3	3	3	2	2	2	3	3	3	2
##	8148	8154	8161	8167	8168	8176	8186	8189	8193	8195	8198	8201	8203
##	3	3	2	1	3	3	1	3	3	3	3	2	3
##	8206	8207	8209	8218	8224	8242	8249	8258	8259	8261	8265	8267	8281
##	3	4	2	4	3	2	3	2	2	1	1	2	2
##	8282	8288	8296	8299	8307	8313	8316	8327	8329	8331	8332	8336	8348
##	3	3	4	2	3	3	2	1	1	3	4	3	2
##	8355	8362	8363	8365	8366	8371	8375	8377	8381	8384	8392	8393	8409
##	4	3	2	2	2	3	2	3	1	3	2	2	3
##	8410	8416	8418	8423	8427	8432	8434	8437	8449	8459	8464	8466	8470
##	3	3	3	4	2	2	1	2	2	2	2	3	2
##	8475	8481	8483	8484	8496	8503	8512	8514	8517	8521	8525	8530	8547
##	3	3	3	3	3	1	3	2	2	3	2	2	4
##	8550	8554	8567	8570	8574	8577	8584	8586	8588	8600	8604	8606	8608
##	3	4	3	2	2	2	2	1	2	2	2	3	4
##	8609	8611	8612	8616	8617	8619	8620	8625	8627	8630	8631	8633	8635
##	2	2	2	2	2	3	2	2	3	2	2	3	2
##	8641	8643	8644	8647	8651	8655	8657	8660	8661	8663	8666	8667	8668
##	3	3	2	3	2	1	3	4	3	2	2	2	3
##	8669	8671	8676	8677	8680	8684	8687	8701	8702	8704	8706	8707	8708
##	3	1	1	3	2	2	3	3	1	1	3	3	2
##	8711	8714	8721	8724	8727	8731	8741	8759	8764	8786	8810	8815	8824
##	2	3	4	3	1	3	3	3	3	1	4	3	3
##	8835	8875	8879	8893	8916	8956	8961	8967	8969	8970	8972	8976	8977
##	3	4	3	2	3	3	3	3	3	3	1	3	3

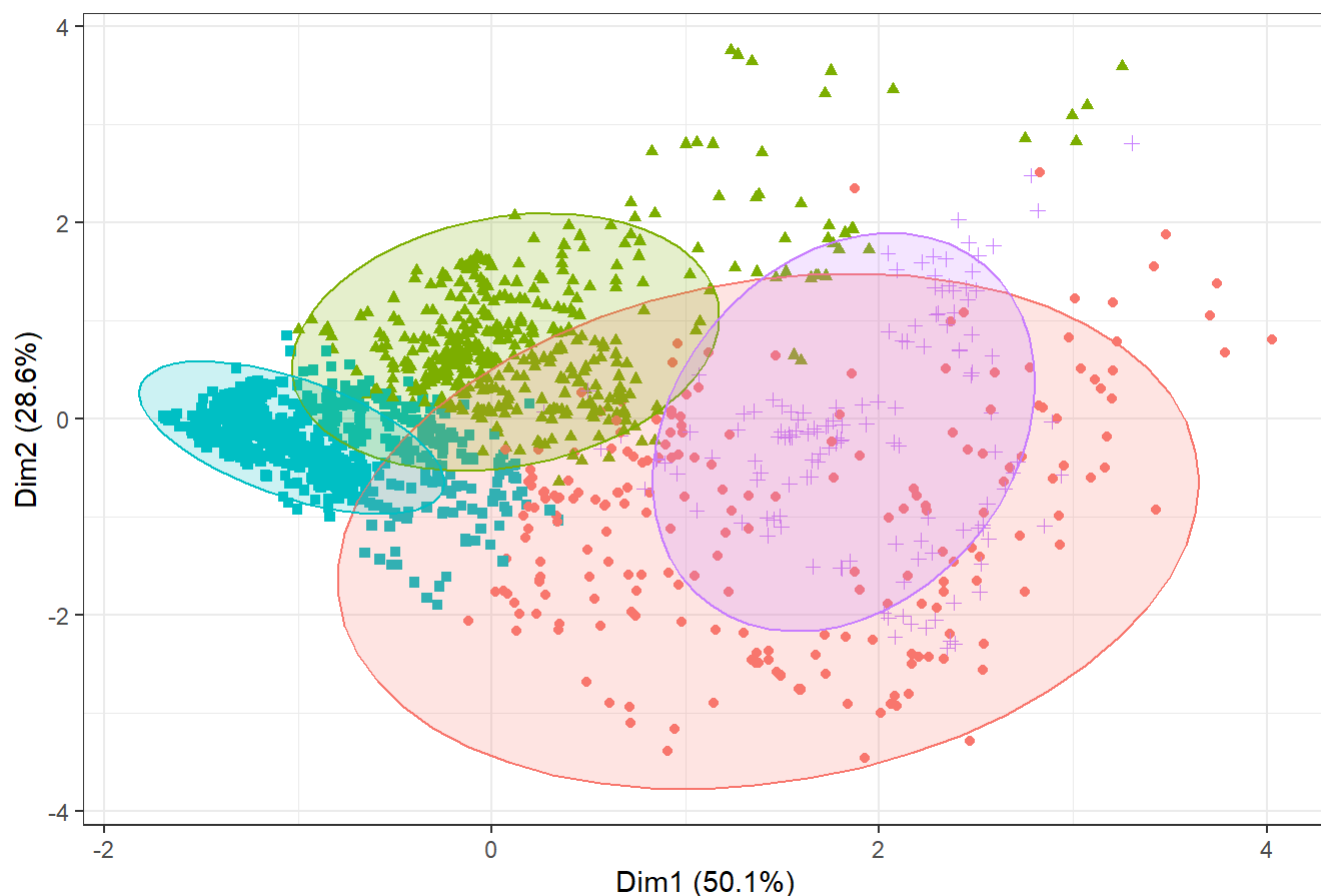
```
## 8978 8981 8983 8984 8986 8991 9002 9003 9004 9007 9008 9016 9022
## 1 3 1 3 1 3 2 3 1 3 1 2 4
## 9025 9027 9029 9031 9034 9036 9040 9045 9048 9049 9052 9054 9055
## 2 3 2 1 3 2 4 2 3 3 3 3 4
## 9057 9060 9064 9070 9071 9072 9078 9089 9091 9093 9096 9100 9108
## 4 3 3 1 3 3 4 4 3 3 1 3 4
## 9110 9112 9116 9117 9119 9120 9129 9133 9136 9145 9153 9158 9173
## 4 1 4 2 3 3 3 4 2 3 3 4 2
## 9179 9180 9189 9195 9197 9201 9204 9216 9218 9226 9227 9229 9237
## 3 3 4 3 3 2 2 3 1 1 3 2 3
## 9238 9247 9260 9261 9262 9268 9271 9279 9282 9288 9295 9297 9300
## 3 3 3 1 1 4 4 2 4 3 4 3 3
## 9303 9304 9308 9318 9320 9335 9339 9340 9342 9349 9350 9353 9377
## 3 3 1 4 3 4 4 3 4 4 3 3 3
## 9387 9389 9390 9392 9394 9397 9406 9422 9427 9439 9449 9451 9453
## 4 3 4 2 1 1 3 4 1 3 4 3 3
## 9458 9477 9483 9487 9488 9507 9526 9532 9534 9536 9538 9541 9549
## 4 1 3 4 2 3 3 1 3 3 4 3 2
## 9555 9559 9561 9571 9572 9578 9579 9597 9598 9605 9609 9613 9619
## 2 3 1 1 4 4 3 1 3 1 3 4 3
## 9625 9634 9637 9640 9641 9643 9649 9655 9658 9669 9673 9675 9679
## 3 3 3 3 1 1 3 3 1 1 3 3 4
## 9683 9684 9685 9689 9711 9727 9752 9762 9775 9777 9783 9785 9798
## 3 3 1 1 3 3 3 1 2 3 2 3 1
## 9807 9828 9831 9834 9836 9843 9858 9860 9866 9867 9873 9875 9878
## 3 3 2 3 2 4 2 4 3 3 3 2 3
## 9880 9881 9887 9890 9892 9905 9909 9936 9943 9946 9952 9953 9965
## 3 2 3 2 4 3 3 4 3 3 1 3 2
## 9970 9972 9975 9991 9994 9996 10000 10003 10004
## 2 3 3 3 1 2 4 1 3
```

```
table(clara_clusters$clustering)
```

```
##
## 1 2 3 4
## 249 611 1074 285
```

```
fviz_cluster(object = clara_clusters, ellipse.type = "t", geom = "point",
              pointsize = 1.5) + theme_bw() +
  labs(title = "Resultados clustering CLARA") +
  theme(legend.position = "none")
```

Resultados clustering CLARA



Seitán

```
clara_clusters <- clara(x = scale(df_seitan[, -1]), k = 3, metric = "manhattan", stand = TRUE,
  samples = 60, pamLike = TRUE)
clara_clusters$sample
```

```
## [1] "473" "816" "824" "1304" "2172" "2207" "2215" "2535" "2952" "3138"
## [11] "3267" "3833" "3838" "3855" "3872" "4212" "4309" "4539" "4789" "5874"
## [21] "5902" "5906" "6144" "6265" "6322" "7085" "7213" "8185" "8196" "8230"
## [31] "8674" "8816" "8818" "8826" "8836" "8922" "8937" "8941" "9104" "9155"
## [41] "9274" "9310" "9370" "9452" "9607" "9801"
```

```
clara_clusters$medoids
```

```
##      PROTEINS_100G CARBOHYDRATES_100G  FAT_100G
## 1304   -0.02248502    -0.2593732 -0.8193032
## 7213    0.77541035    -0.2987133  1.0728682
## 3267   -0.89291633     1.2945632  1.0956655
```

```
clara_clusters$i.med
```

```
## [1] 86 426 177
```

```
clara_clusters$clustering
```

##	16	20	29	32	38	104	105	114	131	133	135	138	147	262	318	368
##	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
##	385	455	471	472	473	490	513	517	533	541	558	568	569	571	572	581
##	2	1	1	1	1	1	1	1	1	3	3	1	1	3	3	2
##	592	597	673	700	719	753	765	809	816	824	828	829	845	857	871	872
##	1	1	2	2	2	3	3	1	1	1	1	1	1	1	1	1
##	891	899	900	902	903	904	905	906	909	910	915	916	918	920	922	923
##	3	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
##	926	933	934	937	939	942	947	948	952	957	982	993	1036	1151	1158	1170
##	1	1	1	1	1	1	1	1	1	3	1	1	1	1	1	3
##	1209	1240	1247	1265	1277	1304	1336	1351	1361	1368	1421	1622	1881	2036	2060	2078
##	3	3	3	3	3	1	1	2	1	2	1	3	3	1	3	1
##	2081	2129	2132	2135	2172	2178	2189	2190	2196	2207	2215	2231	2233	2234	2245	2255
##	1	2	1	3	1	1	2	2	1	2	2	2	2	2	2	1
##	2263	2367	2424	2436	2450	2451	2458	2459	2463	2489	2500	2504	2505	2535	2543	2547
##	2	2	3	2	2	2	2	1	2	2	2	2	2	2	2	2
##	2557	2560	2570	2576	2630	2664	2697	2716	2749	2774	2891	2900	2910	2928	2930	2945
##	2	1	2	3	3	1	3	3	1	1	1	1	1	1	1	3
##	2952	2973	2983	3012	3017	3032	3062	3085	3088	3094	3096	3102	3103	3111	3112	3128
##	2	1	3	1	1	3	1	1	1	1	1	1	3	1	1	1
##	3130	3133	3138	3153	3154	3155	3161	3173	3175	3176	3177	3179	3191	3193	3194	3211
##	1	1	1	3	1	1	2	1	1	2	1	1	3	1	3	1
##	3267	3278	3346	3463	3482	3509	3522	3604	3643	3646	3666	3743	3746	3750	3760	3765
##	3	2	2	3	1	1	1	1	2	1	1	1	3	1	1	1
##	3771	3779	3784	3786	3787	3789	3794	3797	3805	3811	3813	3814	3816	3817	3833	3836
##	1	1	2	1	1	3	1	1	1	1	3	2	1	1	2	1
##	3838	3840	3844	3845	3846	3848	3849	3855	3861	3862	3864	3865	3870	3872	3876	3880
##	3	1	1	1	1	1	1	1	2	1	1	1	3	1	2	1
##	3883	3890	3896	3900	3907	3917	3924	3929	3933	3937	3941	3942	3944	3959	3998	4025
##	1	1	1	1	1	3	1	1	2	1	2	2	1	1	1	1
##	4031	4047	4053	4073	4087	4094	4098	4099	4109	4117	4118	4120	4137	4139	4140	4160
##	1	3	1	2	2	3	1	1	1	3	3	1	3	1	1	1
##	4162	4167	4173	4174	4177	4190	4205	4212	4218	4222	4229	4230	4236	4252	4260	4262
##	1	3	1	3	1	1	1	1	1	1	1	1	3	3	1	1
##	4266	4272	4281	4291	4292	4300	4306	4309	4313	4316	4318	4320	4323	4326	4327	4343
##	1	1	2	1	3	1	1	1	1	3	2	1	3	1	1	2
##	4351	4360	4362	4369	4383	4397	4401	4403	4405	4419	4439	4440	4446	4470	4474	4484
##	3	1	3	3	3	1	3	2	2	3	2	2	3	2	2	1
##	4513	4527	4539	4552	4575	4580	4731	4748	4789	4825	4881	4983	5121	5196	5211	5214
##	1	1	1	1	2	1	1	2	2	3	1	1	1	2	1	1
##	5239	5293	5295	5297	5372	5373	5377	5388	5390	5398	5403	5404	5447	5464	5465	5467
##	1	1	1	1	1	2	2	3	2	1	2	2	1	1	1	1
##	5470	5477	5495	5511	5520	5521	5529	5541	5542	5544	5567	5577	5587	5594	5596	5614
##	2	1	3	1	2	2	2	1	1	2	3	2	1	1	1	3
##	5783	5785	5807	5811	5812	5818	5819	5831	5859	5867	5869	5871	5873	5874	5875	5876
##	1	1	1	1	1	1	1	1	2	1	1	1	1	1	1	1
##	5877	5879	5882	5883	5888	5891	5894	5896	5899	5901	5902	5906	5908	5909	5912	5913
##	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
##	5914	5915	5916	5917	5922	5925	5930	5934	5939	5959	5961	5970	5998	6050	6056	6144
##	1	1	1	1	1	1	1	1	1	1	1	1	1	3	3	1
##	6257	6265	6300	6309	6318	6322	6327	6340	6354	6394	6410	6546	6665	6722	6738	6955
##	1	3	1	3	3	3	3	3	3	1	2	2	1	1	1	1

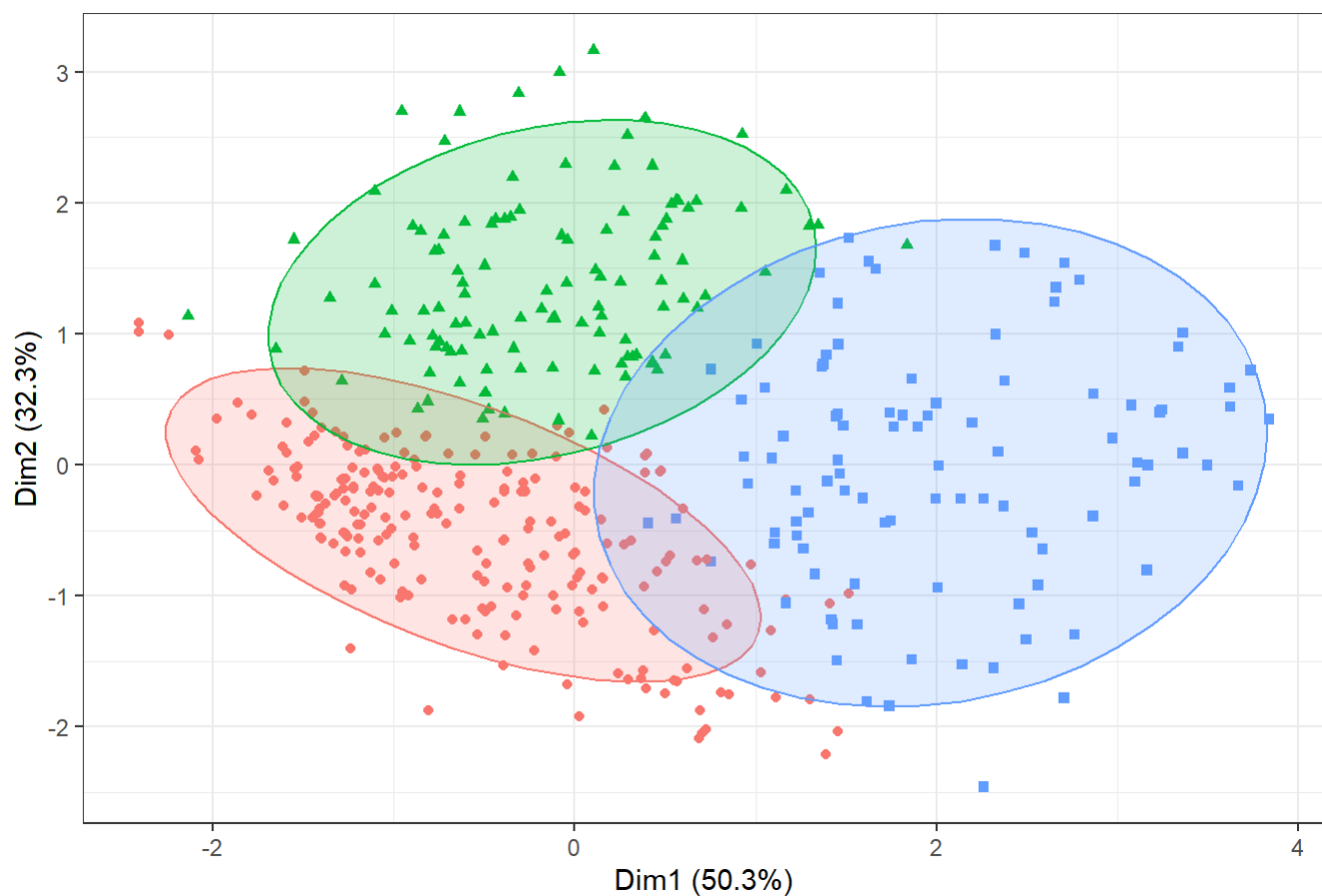

```
## 6971 7032 7085 7104 7114 7184 7195 7198 7208 7213 7215 7227 7243 7246 7254 7265
##    3    3    3    1    1    2    2    1    1    2    1    2    1    1    2    2
## 7268 7275 7302 7306 7330 7341 7376 7424 7476 7503 7519 7529 7533 7535 7536 7542
##    2    2    2    2    3    3    3    3    1    1    2    2    2    2    2    2
## 7551 7573 7577 7583 7600 7631 7642 7644 7645 7654 7663 7665 7679 7684 7685 7735
##    2    2    2    2    2    2    2    2    2    2    2    2    2    2    2    3
## 7766 7806 7931 7952 7960 7962 7990 8029 8052 8057 8070 8084 8101 8160 8170 8180
##    1    2    3    1    1    3    1    1    1    3    1    1    1    1    2    1
## 8181 8184 8185 8190 8196 8197 8199 8221 8222 8226 8230 8231 8233 8234 8251 8270
##    1    1    1    3    1    1    3    1    3    1    1    1    3    1    1    1
## 8302 8310 8317 8400 8599 8622 8653 8659 8674 8710 8723 8736 8738 8744 8745 8747
##    3    1    3    2    3    1    2    1    2    1    1    3    1    3    1    1
## 8750 8753 8767 8774 8779 8780 8787 8793 8796 8805 8806 8816 8818 8825 8826 8829
##    1    3    1    1    2    1    1    1    1    1    1    1    1    1    3    1
## 8832 8833 8836 8838 8840 8841 8842 8845 8848 8850 8851 8853 8854 8860 8874 8877
##    2    2    2    1    2    1    1    1    1    3    1    1    3    1    1    1
## 8889 8895 8901 8904 8906 8922 8924 8927 8937 8941 8943 8950 8952 9046 9058 9085
##    1    1    1    1    1    1    1    3    1    2    2    1    1    2    1    1
## 9104 9125 9141 9146 9154 9155 9156 9162 9163 9178 9181 9182 9188 9191 9192 9217
##    3    3    1    1    3    2    3    1    1    3    1    1    1    1    1    1
## 9222 9234 9236 9249 9257 9258 9274 9283 9292 9293 9296 9309 9310 9326 9344 9345
##    3    1    3    1    1    1    3    2    1    3    3    2    2    1    3    1
## 9355 9370 9393 9402 9404 9407 9409 9411 9429 9436 9446 9450 9452 9464 9468 9475
##    3    3    2    3    1    1    2    3    1    2    2    1    1    2    3    3
## 9495 9497 9501 9506 9516 9521 9527 9528 9563 9580 9583 9593 9607 9616 9621 9678
##    2    3    1    1    2    1    1    1    1    2    2    3    2    2    1    1
## 9773 9789 9800 9801 9833
##    1    1    2    2    2
```

```
table(clara_clusters$clustering)
```

```
##
##    1    2    3
## 377 137 115
```

```
fviz_cluster(object = clara_clusters, ellipse.type = "t", geom = "point",
              pointsize = 1.5) + theme_bw() +
  labs(title = "Resultados clustering CLARA") +
  theme(legend.position = "none")
```

Resultados clustering CLARA



Tofu

```
df_tofu <- na.omit(df_tofu)
```

```
clara_clusters <- clara(x = scale(df_tofu[,-1]), k = 3, metric = "manhattan", stand = TRUE,
                        samples = 60, pamLike = TRUE)
clara_clusters$sample
```

```
## [1] "34"  "218" "275" "506" "601" "1431" "1943" "1955" "1984" "2155"
## [11] "2194" "2510" "2741" "3547" "3559" "3717" "3881" "3940" "4679" "5034"
## [21] "5084" "5207" "5222" "5228" "5397" "5473" "5550" "5551" "5633" "5711"
## [31] "5803" "6975" "6995" "7197" "7280" "7368" "7612" "7658" "7973" "8121"
## [41] "8404" "8856" "8938" "9177" "9248" "9502"
```

```
clara_clusters$medoids
```

```
##          PROTEINS_100G CARBOHYDRATES_100G      FAT_100G
## 1431      0.6526350      -0.6222919  0.09686495
## 5711     -0.9897297       1.8411453 -0.06455252
## 5397     -0.7189840      -0.4994559 -0.80110451
```

```
clara_clusters$i.med
```

```
## [1] 461 1553 1428
```

```
clara_clusters$clustering
```

##	3	4	6	7	8	13	18	19	22	31	33	34	35
##	1	2	1	2	1	3	3	1	1	1	2	1	3
##	41	42	43	48	49	51	54	55	57	58	59	60	61
##	3	3	3	3	3	2	1	1	2	2	1	3	2
##	62	63	65	66	67	68	69	70	71	72	74	76	77
##	3	1	3	3	3	1	3	3	3	1	3	3	3
##	81	84	85	86	87	88	89	91	92	93	94	96	98
##	1	2	3	3	3	3	1	1	1	3	3	1	3
##	100	102	106	107	108	109	110	112	113	115	116	117	118
##	3	3	3	1	1	2	3	1	3	1	2	2	3
##	119	120	122	123	124	125	126	127	128	129	132	134	137
##	3	3	2	1	2	2	2	3	3	3	3	1	3
##	139	141	142	143	144	146	149	152	153	155	156	158	159
##	1	3	2	2	3	1	2	3	3	1	2	3	1
##	163	164	166	167	168	171	173	175	177	178	180	184	187
##	1	1	3	1	3	1	3	1	3	1	3	3	1
##	189	190	191	192	193	194	195	196	197	199	201	202	205
##	1	3	2	2	1	3	3	1	3	1	1	1	1
##	206	208	209	212	215	216	217	218	220	221	222	224	225
##	1	1	1	1	1	2	1	1	1	3	1	1	1
##	228	229	231	234	236	237	238	241	242	243	244	245	248
##	1	1	3	3	3	3	3	3	3	3	3	3	3
##	250	251	252	254	256	257	259	260	263	264	265	267	268
##	1	3	3	3	3	3	1	1	3	3	3	1	2
##	269	270	271	272	274	275	276	277	278	283	290	297	302
##	3	3	3	1	3	3	3	3	3	3	3	3	3
##	303	306	310	311	313	315	316	320	321	322	323	324	330
##	3	3	2	3	3	3	3	2	1	3	3	3	2
##	333	334	335	336	350	351	353	355	356	358	361	363	369
##	3	3	3	3	3	2	3	3	1	3	3	1	1
##	382	383	390	392	394	397	399	402	409	414	418	423	424
##	2	2	3	3	1	3	2	1	1	3	3	1	1
##	425	430	431	434	435	436	445	448	453	456	457	460	461
##	3	2	1	3	1	3	2	2	1	3	1	3	3
##	465	467	469	478	483	488	489	492	495	496	499	500	503
##	3	2	1	3	3	3	3	1	3	1	2	1	3
##	505	506	509	510	511	516	518	522	525	526	527	528	530
##	3	3	3	1	3	2	1	2	1	3	3	3	3
##	534	535	536	537	538	543	544	546	548	549	554	555	559
##	3	1	3	3	2	1	2	1	2	3	1	1	1
##	560	562	565	572	576	577	579	580	583	584	587	589	590
##	3	1	2	2	1	3	3	2	3	3	1	3	1
##	595	596	598	600	601	603	604	605	606	608	609	610	614
##	2	3	1	3	1	1	3	3	2	2	3	2	2
##	618	619	620	621	623	624	626	627	628	630	631	633	634
##	2	1	2	1	3	1	2	3	3	2	3	3	3
##	635	636	637	638	645	650	651	653	655	656	657	660	661
##	2	3	1	1	2	3	1	1	1	1	1	2	1
##	668	670	671	674	676	679	680	682	685	691	692	693	697
##	3	2	3	1	2	3	2	3	1	1	3	3	1
##	698	699	700	703	718	721	726	733	736	738	740	746	747
##	3	3	1	3	1	1	1	3	3	1	1	3	1

##	752	755	756	757	761	769	773	782	785	792	795	800	801
##	3	3	1	1	2	3	3	1	3	3	1	1	1
##	802	804	806	807	808	810	813	818	819	821	826	827	830
##	1	2	1	1	1	1	1	1	1	1	1	3	1
##	831	838	842	846	848	862	864	866	870	874	878	883	885
##	1	2	2	1	3	3	1	2	1	3	1	1	1
##	888	889	890	893	907	914	921	928	936	943	945	949	953
##	1	1	1	1	1	2	3	1	3	1	1	1	1
##	954	958	960	961	964	969	972	974	976	986	988	989	990
##	1	1	1	3	1	1	3	1	1	1	1	1	1
##	1006	1011	1016	1034	1038	1045	1048	1052	1053	1061	1063	1065	1069
##	1	1	1	1	1	1	2	3	1	2	1	1	2
##	1071	1075	1076	1082	1087	1088	1090	1093	1109	1120	1123	1125	1127
##	2	1	2	3	3	2	1	1	3	2	3	1	1
##	1157	1174	1177	1183	1205	1207	1213	1217	1218	1223	1230	1236	1241
##	3	3	1	2	3	2	1	1	1	3	2	1	2
##	1242	1243	1244	1258	1260	1269	1273	1274	1280	1287	1290	1300	1309
##	1	1	3	3	3	3	1	2	1	1	1	1	1
##	1310	1313	1317	1397	1427	1431	1443	1468	1494	1534	1538	1541	1542
##	2	1	1	1	2	1	2	3	1	2	3	1	1
##	1550	1572	1578	1598	1599	1600	1602	1605	1606	1612	1613	1614	1617
##	1	1	3	1	1	1	2	3	1	3	2	2	1
##	1621	1633	1637	1638	1643	1646	1648	1651	1653	1655	1657	1658	1663
##	1	1	2	2	1	3	1	1	1	1	2	1	2
##	1664	1670	1671	1672	1673	1678	1702	1720	1735	1766	1771	1772	1777
##	1	1	1	1	2	1	3	2	1	2	1	2	1
##	1807	1808	1815	1825	1832	1837	1838	1841	1847	1848	1850	1852	1858
##	1	1	1	1	1	1	1	1	2	2	1	2	3
##	1861	1863	1867	1874	1890	1897	1912	1915	1919	1923	1925	1929	1936
##	2	3	2	2	2	2	1	1	1	3	3	1	3
##	1943	1944	1947	1954	1955	1958	1959	1960	1968	1969	1970	1973	1974
##	2	3	2	1	2	3	1	1	2	3	1	1	2
##	1982	1983	1984	1993	1995	2008	2018	2023	2027	2032	2039	2043	2045
##	3	3	2	3	3	1	3	3	2	2	1	1	3
##	2049	2056	2063	2067	2072	2108	2111	2112	2117	2120	2121	2122	2123
##	1	3	3	1	1	1	1	1	1	1	1	1	2
##	2126	2127	2134	2136	2138	2139	2140	2144	2146	2149	2152	2154	2155
##	1	1	3	1	1	1	1	3	1	1	1	2	1
##	2156	2157	2159	2160	2163	2164	2166	2167	2169	2170	2171	2173	2174
##	1	1	1	2	1	1	1	2	1	1	2	1	1
##	2176	2179	2180	2183	2186	2187	2188	2191	2194	2198	2199	2200	2202
##	2	3	2	2	1	1	1	1	3	1	1	1	1
##	2205	2206	2210	2214	2221	2224	2228	2230	2232	2238	2240	2241	2243
##	2	1	1	1	1	2	1	1	1	1	1	1	1
##	2244	2246	2248	2249	2250	2251	2253	2259	2265	2272	2276	2280	2284
##	1	3	2	1	1	3	1	1	1	1	1	1	1
##	2285	2290	2291	2298	2301	2302	2304	2306	2310	2315	2317	2318	2321
##	2	1	1	1	1	3	3	1	2	1	1	1	1
##	2322	2323	2327	2329	2336	2339	2342	2343	2347	2349	2352	2353	2355
##	3	1	1	1	1	3	1	1	1	1	1	1	1
##	2357	2359	2365	2372	2379	2380	2382	2388	2389	2390	2393	2394	2397
##	1	3	1	1	1	1	1	1	1	1	1	1	1

##	2402	2404	2405	2406	2407	2408	2411	2412	2414	2419	2420	2421	2433
##	1	1	1	1	1	1	1	1	1	1	1	1	1
##	2435	2442	2443	2445	2447	2461	2470	2472	2481	2482	2499	2501	2506
##	1	1	1	1	1	1	3	1	3	1	1	1	1
##	2508	2510	2514	2515	2516	2519	2528	2530	2538	2540	2549	2554	2561
##	1	1	1	1	1	1	1	1	1	1	1	3	3
##	2569	2573	2576	2578	2583	2586	2590	2591	2592	2598	2599	2601	2602
##	1	1	1	1	2	1	3	1	1	1	1	1	2
##	2606	2612	2613	2616	2621	2626	2628	2640	2650	2656	2665	2673	2676
##	1	1	1	1	3	2	2	3	2	1	3	3	3
##	2692	2700	2707	2708	2709	2712	2723	2729	2737	2739	2741	2746	2750
##	1	3	1	1	2	3	3	1	2	3	2	1	3
##	2752	2753	2754	2760	2763	2768	2769	2773	2778	2780	2786	2790	2803
##	1	2	1	1	1	1	1	1	1	2	2	1	3
##	2804	2805	2810	2811	2814	2816	2819	2825	2826	2829	2834	2836	2839
##	2	1	1	1	2	1	1	3	1	2	1	2	3
##	2847	2863	2864	2867	2886	2899	2901	2941	2949	2969	2976	2979	3004
##	1	3	2	1	1	3	3	1	1	1	1	3	1
##	3021	3025	3061	3074	3076	3087	3104	3105	3106	3122	3129	3137	3140
##	1	1	1	1	3	1	1	3	1	1	3	1	1
##	3152	3172	3174	3202	3204	3205	3206	3208	3209	3214	3222	3223	3224
##	3	1	3	1	1	1	1	1	3	3	1	1	1
##	3227	3229	3234	3236	3238	3241	3250	3251	3262	3264	3271	3272	3277
##	1	1	1	1	1	1	1	1	2	1	3	2	1
##	3294	3297	3306	3308	3309	3316	3327	3329	3332	3335	3340	3342	3360
##	1	1	1	1	1	1	1	2	3	1	1	1	1
##	3366	3375	3390	3401	3405	3422	3430	3431	3434	3441	3447	3451	3454
##	1	1	3	1	1	1	1	1	2	3	3	2	3
##	3455	3458	3459	3460	3471	3474	3484	3495	3496	3499	3515	3528	3530
##	2	1	3	1	2	2	1	1	1	1	3	1	2
##	3531	3536	3539	3545	3547	3548	3549	3551	3552	3555	3556	3557	3558
##	1	1	1	3	2	1	2	1	1	1	1	1	1
##	3559	3561	3566	3568	3570	3578	3580	3581	3582	3583	3585	3589	3590
##	2	2	1	1	2	1	3	1	1	3	2	1	2
##	3591	3593	3594	3595	3597	3599	3606	3607	3615	3616	3617	3618	3621
##	3	3	1	1	2	2	3	1	1	1	1	1	1
##	3627	3629	3633	3636	3645	3647	3648	3669	3671	3674	3683	3702	3703
##	2	3	1	3	1	1	1	3	1	1	1	1	1
##	3717	3719	3723	3745	3748	3749	3752	3753	3755	3758	3761	3772	3780
##	1	3	1	3	2	2	1	1	2	1	2	1	1
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##	1	1	1	1	1	2	1	3	3	1	1	3	1
##	3850	3851	3852	3853	3854	3856	3857	3858	3859	3866	3867	3868	3871
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##	3877	3879	3881	3884	3885	3886	3887	3891	3892	3893	3894	3895	3897
##	1	1	1	1	1	1	2	1	1	1	1	1	1
##	3898	3899	3901	3904	3906	3908	3910	3914	3922	3923	3926	3931	3940
##	1	3	1	2	1	1	3	1	2	1	1	2	3
##	3941	3942	3950	3955	3962	3973	4002	4009	4017	4023	4027	4052	4058
##	1	1	3	3	1	1	1	1	3	3	3	3	3

##	4076	4079	4089	4090	4091	4093	4102	4104	4105	4107	4108	4110	4112
##	3	1	3	1	1	1	1	1	1	1	2	1	1
##	4116	4124	4125	4128	4130	4133	4134	4136	4142	4146	4154	4157	4161
##	1	2	1	3	1	3	2	2	1	2	3	3	1
##	4163	4165	4166	4174	4175	4179	4181	4183	4185	4188	4191	4192	4193
##	1	1	2	2	3	2	2	2	2	3	3	3	1
##	4195	4198	4204	4211	4213	4216	4217	4231	4235	4242	4243	4254	4255
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##	5547	5550	5551	5552	5554	5556	5557	5558	5560	5562	5564	5565	5566
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##	1	2	2	1	1	3	1	1	1	1	2	1	2

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## 9512 9514 9517 9530 9535 9539 9545 9548 9550 9551 9557 9562 9564
##      2      3      1      1      1      2      1      3      2      1      1      1      3
## 9568 9569 9574 9576 9585 9589 9590 9592 9603 9604 9612 9617 9644
##      3      3      3      1      1      1      1      1      2      1      1      1      1
## 9656 9666 9668 9681 9692 9703 9707 9709 9715 9718 9719 9726 9732
##      1      1      1      1      3      3      3      1      2      3      2      3      1
## 9735 9738 9748 9755 9759 9763 9766 9768 9776 9779 9780 9788 9792
##      3      1      2      1      3      1      1      1      1      1      2      3      1
## 9795 9797 9802 9803 9814 9815 9818 9826 9827 9829 9830 9836 9837
##      3      3      3      1      3      2      1      1      3      1      3      1      3
## 9838 9840 9842 9844 9845 9846 9848 9851 9853 9854 9856 9857 9859
##      1      1      2      2      3      1      3      2      3      1      1      1      3
## 9861 9864 9865 9868 9869 9870 9874 9879 9883 9884 9885 9888 9889
##      1      1      3      1      2      3      2      3      1      3      2      1      1
## 9894 9898 9899 9903 9910 9914 9920 9922 9923 9924 9925 9926 9928
##      1      3      3      1      3      2      1      2      2      1      3      1      1
## 9929 9931 9932 9935 9937 9938 9942 9950 9954 9957 9958 9959 9960
##      3      1      1      1      3      3      1      2      3      1      1      3      3
## 9961 9962 9964 9966 9967 9974 9978 9980 9981 9986 9993 10002 10008
##      3      1      2      1      1      2      3      2      3      1      3      1      2
```

```
table(clara_clusters$clustering)
```

```
##
##      1      2      3
## 1349  458  702
```

```
fviz_cluster(object = clara_clusters, ellipse.type = "t", geom = "point",
              pointsize = 1.5) + theme_bw() +
  labs(title = "Resultados clustering CLARA") +
  theme(legend.position = "none")
```

Resultados clustering CLARA

