

Anàlisi de dades òmiques (M0-157). PAC1.

Marc Anton

2025-03-26

1. Introducció

Aquest arxiu conté les sol·lucions als problemes plantejats en la PAC1 *Anàlisi de dades òmiques (M0-157)*. El projecte total del treball es pot consultar al: [meu repositori github](#), cliqueu **aquí**

2. Carrega de dades

A continuació, procedeix a resoldre la PAC. Abans que res, carreguem les dades al nostre entorn de treball. Treballarem amb l'exemple de mostra, veure **aquí**. Carreguem els dos fitxers:

```
DataInfo_S013 <- read.csv("DataInfo_S013.csv", sep = ",")
DataValues_S013 <- read.csv("DataValues_S013.csv", sep = ",")
```

Com que ja sabem que aquests fitxers solen portar molts camps, mirem primer l'estructura dels dos.

3. Exploració previa de les dades descarregades

```
dim(DataInfo_S013)
```

```
## [1] 695 4
```

```
dim(DataValues_S013)
```

```
## [1] 39 696
```

Aquí ja veiem que en l'arxiu Values tenim 696 columnes, el que correspon a un index únic de cada cas d'estudi més les 695 variables que es tenen de cada un d'ells. Per veure que hi ha, mirem les 20 primeres files d'Info i podem fer-nos una idea

```
head(DataInfo_S013,20)
```

##	X	VarName	varTpe	Description
## 1	SUBJECTS	SUBJECTS	integer	dataDesc
## 2	SURGERY	SURGERY	character	dataDesc
## 3	AGE	AGE	integer	dataDesc
## 4	GENDER	GENDER	character	dataDesc
## 5	Group	Group	integer	dataDesc
## 6	MEDDM_T0	MEDDM_T0	integer	dataDesc
## 7	MEDCOL_T0	MEDCOL_T0	integer	dataDesc
## 8	MEDINF_T0	MEDINF_T0	integer	dataDesc
## 9	MEDHTA_T0	MEDHTA_T0	integer	dataDesc
## 10	GLU_T0	GLU_T0	integer	dataDesc
## 11	INS_T0	INS_T0	numeric	dataDesc

```
## 12          HOMA_TO          HOMA_TO    numeric    dataDesc
## 13          HBA1C_TO          HBA1C_TO    numeric    dataDesc
## 14 HBA1C.mmol.mol_TO HBA1C.mmol.mol_TO    numeric    dataDesc
## 15          PESO_TO          PESO_TO     integer    dataDesc
## 16          bmi_TO          bmi_TO      numeric    dataDesc
## 17          CC_TO          CC_TO      numeric    dataDesc
## 18          CINT_TO          CINT_TO     integer    dataDesc
## 19          CAD_TO          CAD_TO      integer    dataDesc
## 20          TAD_TO          TAD_TO      integer    dataDesc
```

Aquí ja veiem que hi ha 9 primers camps que semblen ser els que tenen la informació sobre el pacient mostra. Mirem que contenen aquests camps a values:

```
# En mirem 10 en realitat, per comprovar que la primera columna és un index
head(DataValues_S013[,1:10])
```

```
##   X.1 SUBJECTS SURGERY AGE GENDER Group MEDDM_TO MEDCOL_TO MEDINF_TO MEDHTA_TO
## 1   1         1 by pass  27     F     1         0         0         0         1
## 2   2         2 by pass  19     F     2         0         0         0         0
## 3   3         3 by pass  42     F     1         0         0         0         0
## 4   4         4 by pass  37     F     2         0         0         0         0
## 5   5         5 tubular  42     F     1         0         0         0         0
## 6   6         6 by pass  24     F     2         0         0         0         0
```

Com suposavem, aquestes columnes corresponen a: 1. Un codi de pacient (que està repetit a les dues primeres columnes) 2. La cirurgia a la que va ser sotmès el pacient 3. L'edat i el gènere 4. Un grup de tractament 5. 4 tractaments que venen amb 0 i 1

Per veure que la resta són metabolits, mirem 5 columnes més

```
head(DataValues_S013[,11:15],10)
```

```
##      GLU_TO INS_TO HOMA_TO HBA1C_TO HBA1C.mmol.mol_TO
## 1      85  11.40   2.40      NA              NA
## 2      78  12.10   2.32      NA              NA
## 3      75   8.41   1.56     5.4          35.51
## 4      71  12.80   2.25     5.1          32.23
## 5      82   6.01   1.22     5.6          37.69
## 6      71   9.88   1.73     5.1          32.23
## 7      80   9.20   1.82     5.6          37.69
## 8      90   3.40   0.76     5.5          36.60
## 9      92   5.43   1.23     5.7          38.78
## 10     84   6.98   1.45     5.5          36.60
```

Obviament ja ho veiem i també podem veure que hi ha més d'un valor que no es té (NA)

4. Generació de l'object SummarizedExperiment

Ara que ja sabem que tenim a les dades, podem procedir a carregar-les en un objecte *SummarizedExperiment*. Per a fer-ho, convertirem però les dades a un data frame i li traurem la primera columna que ja hem vist que és redundant amb la segona. Aleshores ja ho podem carregar. En concret, carregarem:

1. L'arxiu DataValues com a assays que és on es guarden les dades de les mostres que tenim
2. L'arxiu DataInfo com a colData que és on es guarda la informació de com és cada columna d'assays.

```
# Primer carreguem la llibreria per poder tractar el SummarizedExperiment
library(SummarizedExperiment)
```

```

## Loading required package: MatrixGenerics
## Loading required package: matrixStats
##
## Attaching package: 'MatrixGenerics'
## The following objects are masked from 'package:matrixStats':
##
##   colAlls, colAnyNAs, colAnys, colAvgPerRowSet, colCollapse,
##   colCounts, colCummaxs, colCummins, colCumprods, colCumsums,
##   colDiffs, colIQRDiffs, colIQRs, colLogSumExps, colMadDiffs,
##   colMads, colMaxs, colMeans2, colMedians, colMins, colOrderStats,
##   colProds, colQuantiles, colRanges, colRanks, colSdDiffs, colSds,
##   colSums2, colTabulates, colVarDiffs, colVars, colWeightedMads,
##   colWeightedMeans, colWeightedMedians, colWeightedSds,
##   colWeightedVars, rowAlls, rowAnyNAs, rowAnys, rowAvgPerColSet,
##   rowCollapse, rowCounts, rowCummaxs, rowCummins, rowCumprods,
##   rowCumsums, rowDiffs, rowIQRDiffs, rowIQRs, rowLogSumExps,
##   rowMadDiffs, rowMads, rowMaxs, rowMeans2, rowMedians, rowMins,
##   rowOrderStats, rowProds, rowQuantiles, rowRanges, rowRanks,
##   rowSdDiffs, rowSds, rowSums2, rowTabulates, rowVarDiffs, rowVars,
##   rowWeightedMads, rowWeightedMeans, rowWeightedMedians,
##   rowWeightedSds, rowWeightedVars
## Loading required package: GenomicRanges
## Loading required package: stats4
## Loading required package: BiocGenerics
##
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:stats':
##
##   IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
##   anyDuplicated, aperm, 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, saveRDS, setdiff,
##   table, tapply, union, unique, unsplit, which.max, which.min
## Loading required package: S4Vectors
##
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:utils':
##
##   findMatches
## The following objects are masked from 'package:base':
##
##   expand.grid, I, unname
## Loading required package: IRanges

```

```
##
## Attaching package: 'IRanges'

## The following object is masked from 'package:grDevices':
##
##     windows

## Loading required package: GenomeInfoDb

## 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)".

##
## Attaching package: 'Biobase'

## The following object is masked from 'package:MatrixGenerics':
##
##     rowMedians

## The following objects are masked from 'package:matrixStats':
##
##     anyMissing, rowMedians

# Primer eliminem la primera columna de DataValues i ho guardem com un data frame que és el que accepta
DataValues_S013_matrix <- DataValues_S013[,-1]
# Ara ja podem carregar el SummarizedExperiment
seMA <- SummarizedExperiment(assays = list(counts = DataValues_S013_matrix),
                             colData = DataInfo_S013)
seMA

## class: SummarizedExperiment
## dim: 39 695
## metadata(0):
## assays(1): counts
## rownames(39): 1 2 ... 38 39
## rowData names(0):
## colnames(695): SUBJECTS SURGERY ... SM.C24.0_T5 SM.C24.1_T5
## colData names(4): X VarName varTpe Description

Sembla que l'objecte s'ha generat correctament

Mirem un parell de funcions que es poden fer.

# Comprovem com es diuen les columnes que tenim en cada una de les classes de dins de l'objecte
dimnames(seMA)

## [[1]]
## [1] "1" "2" "3" "4" "5" "6" "7" "8" "9" "10" "11" "12" "13" "14" "15"
## [16] "16" "17" "18" "19" "20" "21" "22" "23" "24" "25" "26" "27" "28" "29" "30"
## [31] "31" "32" "33" "34" "35" "36" "37" "38" "39"
##
## [[2]]
## [1] "SUBJECTS" "SURGERY" "AGE"
## [4] "GENDER" "Group" "MEDDM_TO"
## [7] "MEDCOL_TO" "MEDINF_TO" "MEDHTA_TO"
```

##	[10]	"GLU_T0"	"INS_T0"	"HOMA_T0"
##	[13]	"HBA1C_T0"	"HBA1C.mmol.mol_T0"	"PESO_T0"
##	[16]	"bmi_T0"	"CC_T0"	"CINT_T0"
##	[19]	"CAD_T0"	"TAD_T0"	"TAS_T0"
##	[22]	"TG_T0"	"COL_T0"	"LDL_T0"
##	[25]	"HDL_T0"	"VLDL_T0"	"PCR_T0"
##	[28]	"LEP_T0"	"ADIPO_T0"	"GOT_T0"
##	[31]	"GPT_T0"	"GGT_T0"	"URICO_T0"
##	[34]	"CREAT_T0"	"UREA_T0"	"HIERRO_T0"
##	[37]	"TRANSF_T0"	"FERR_T0"	"Ile_T0"
##	[40]	"Leu_T0"	"Val_T0"	"Ala_T0"
##	[43]	"Pro_T0"	"Gly_T0"	"Ser_T0"
##	[46]	"Trp_T0"	"Phe_T0"	"Met_T0"
##	[49]	"Orn_T0"	"Arg_T0"	"His_T0"
##	[52]	"Asn_T0"	"Asp_T0"	"Glu_T0"
##	[55]	"Gln_T0"	"Cit_T0"	"Tyr_T0"
##	[58]	"Thr_T0"	"Lys_T0"	"Creatinine_T0"
##	[61]	"Kynurenine_T0"	"Putrescine_T0"	"Sarcosine_T0"
##	[64]	"Serotonin_T0"	"Taurine_T0"	"SDMA_T0"
##	[67]	"C0_T0"	"C2_T0"	"C3.OH_T0"
##	[70]	"C6..C4.1.DC._T0"	"C5.DC..C6.OH._T0"	"C7.DC_T0"
##	[73]	"C8_T0"	"C10_T0"	"C10.1_T0"
##	[76]	"C10.2_T0"	"C14.1_T0"	"C14.2_T0"
##	[79]	"C16.1_T0"	"C16.2_T0"	"C16.2.OH_T0"
##	[82]	"C18.1_T0"	"C18.1.OH_T0"	"C18.2_T0"
##	[85]	"lysoPC.a.C16.0_T0"	"lysoPC.a.C16.1_T0"	"lysoPC.a.C17.0_T0"
##	[88]	"lysoPC.a.C18.0_T0"	"lysoPC.a.C18.1_T0"	"lysoPC.a.C18.2_T0"
##	[91]	"lysoPC.a.C20.3_T0"	"lysoPC.a.C20.4_T0"	"lysoPC.a.C24.0_T0"
##	[94]	"lysoPC.a.C26.0_T0"	"lysoPC.a.C26.1_T0"	"lysoPC.a.C28.0_T0"
##	[97]	"lysoPC.a.C28.1_T0"	"PC.aa.C24.0_T0"	"PC.aa.C28.1_T0"
##	[100]	"PC.aa.C30.0_T0"	"PC.aa.C32.0_T0"	"PC.aa.C32.1_T0"
##	[103]	"PC.aa.C32.3_T0"	"PC.aa.C34.1_T0"	"PC.aa.C34.2_T0"
##	[106]	"PC.aa.C34.3_T0"	"PC.aa.C34.4_T0"	"PC.aa.C36.0_T0"
##	[109]	"PC.aa.C36.1_T0"	"PC.aa.C36.2_T0"	"PC.aa.C36.3_T0"
##	[112]	"PC.aa.C36.4_T0"	"PC.aa.C36.5_T0"	"PC.aa.C38.0_T0"
##	[115]	"PC.aa.C38.1_T0"	"PC.aa.C38.3_T0"	"PC.aa.C38.4_T0"
##	[118]	"PC.aa.C38.5_T0"	"PC.aa.C38.6_T0"	"PC.aa.C40.1_T0"
##	[121]	"PC.aa.C40.2_T0"	"PC.aa.C40.3_T0"	"PC.aa.C40.4_T0"
##	[124]	"PC.aa.C40.5_T0"	"PC.aa.C40.6_T0"	"PC.aa.C42.0_T0"
##	[127]	"PC.aa.C42.1_T0"	"PC.aa.C42.2_T0"	"PC.aa.C42.4_T0"
##	[130]	"PC.aa.C42.5_T0"	"PC.aa.C42.6_T0"	"PC.aa.C30.0_T0"
##	[133]	"PC.aa.C32.1_T0"	"PC.aa.C32.2_T0"	"PC.aa.C34.0_T0"
##	[136]	"PC.aa.C34.1_T0"	"PC.aa.C34.2_T0"	"PC.aa.C34.3_T0"
##	[139]	"PC.aa.C36.0_T0"	"PC.aa.C36.1_T0"	"PC.aa.C36.2_T0"
##	[142]	"PC.aa.C36.3_T0"	"PC.aa.C36.4_T0"	"PC.aa.C36.5_T0"
##	[145]	"PC.aa.C38.0_T0"	"PC.aa.C38.2_T0"	"PC.aa.C38.3_T0"
##	[148]	"PC.aa.C38.4_T0"	"PC.aa.C38.5_T0"	"PC.aa.C38.6_T0"
##	[151]	"PC.aa.C40.1_T0"	"PC.aa.C40.2_T0"	"PC.aa.C40.3_T0"
##	[154]	"PC.aa.C40.4_T0"	"PC.aa.C40.5_T0"	"PC.aa.C40.6_T0"
##	[157]	"PC.aa.C42.1_T0"	"PC.aa.C42.2_T0"	"PC.aa.C42.3_T0"
##	[160]	"PC.aa.C42.4_T0"	"PC.aa.C42.5_T0"	"PC.aa.C44.3_T0"
##	[163]	"PC.aa.C44.4_T0"	"PC.aa.C44.5_T0"	"PC.aa.C44.6_T0"
##	[166]	"SM..OH..C14.1_T0"	"SM..OH..C16.1_T0"	"SM..OH..C22.1_T0"
##	[169]	"SM..OH..C22.2_T0"	"SM..OH..C24.1_T0"	"SM.C16.0_T0"

## [172]	"SM.C16.1_T0"	"SM.C18.0_T0"	"SM.C18.1_T0"
## [175]	"SM.C20.2_T0"	"SM.C24.0_T0"	"SM.C24.1_T0"
## [178]	"MEDDM_T2"	"MEDCOL_T2"	"MEDINF_T2"
## [181]	"MEDHTA_T2"	"GLU_T2"	"INS_T2"
## [184]	"HOMA_T2"	"HBA1C_T2"	"HBA1C.mmol.mol_T2"
## [187]	"PESO_T2"	"bmi_T2"	"CC_T2"
## [190]	"CINT_T2"	"CAD_T2"	"TAD_T2"
## [193]	"TAS_T2"	"TG_T2"	"COL_T2"
## [196]	"LDL_T2"	"HDL_T2"	"VLDL_T2"
## [199]	"PCR_T2"	"LEP_T2"	"ADIPO_T2"
## [202]	"GOT_T2"	"GPT_T2"	"GGT_T2"
## [205]	"URICO_T2"	"CREAT_T2"	"UREA_T2"
## [208]	"HIERRO_T2"	"TRANSF_T2"	"FERR_T2"
## [211]	"Ile_T2"	"Leu_T2"	"Val_T2"
## [214]	"Ala_T2"	"Pro_T2"	"Gly_T2"
## [217]	"Ser_T2"	"Trp_T2"	"Phe_T2"
## [220]	"Met_T2"	"Orn_T2"	"Arg_T2"
## [223]	"His_T2"	"Asn_T2"	"Asp_T2"
## [226]	"Glu_T2"	"Gln_T2"	"Cit_T2"
## [229]	"Tyr_T2"	"Thr_T2"	"Lys_T2"
## [232]	"Creatinine_T2"	"Kynurenine_T2"	"Putrescine_T2"
## [235]	"Sarcosine_T2"	"Serotonin_T2"	"Taurine_T2"
## [238]	"SDMA_T2"	"CO_T2"	"C2_T2"
## [241]	"C3.OH_T2"	"C6..C4.1.DC._T2"	"C5.DC..C6.OH._T2"
## [244]	"C7.DC_T2"	"C8_T2"	"C10_T2"
## [247]	"C10.1_T2"	"C10.2_T2"	"C14.1_T2"
## [250]	"C14.2_T2"	"C16.1_T2"	"C16.2_T2"
## [253]	"C16.2.OH_T2"	"C18.1_T2"	"C18.1.OH_T2"
## [256]	"C18.2_T2"	"X"	"lysoPC.a.C14.0_T2"
## [259]	"lysoPC.a.C16.0_T2"	"lysoPC.a.C16.1_T2"	"lysoPC.a.C17.0_T2"
## [262]	"lysoPC.a.C18.0_T2"	"lysoPC.a.C18.1_T2"	"lysoPC.a.C18.2_T2"
## [265]	"lysoPC.a.C20.3_T2"	"lysoPC.a.C20.4_T2"	"lysoPC.a.C24.0_T2"
## [268]	"lysoPC.a.C26.0_T2"	"lysoPC.a.C26.1_T2"	"lysoPC.a.C28.0_T2"
## [271]	"lysoPC.a.C28.1_T2"	"PC.aa.C24.0_T2"	"PC.aa.C28.1_T2"
## [274]	"PC.aa.C30.0_T2"	"PC.aa.C32.0_T2"	"PC.aa.C32.1_T2"
## [277]	"PC.aa.C32.3_T2"	"PC.aa.C34.1_T2"	"PC.aa.C34.2_T2"
## [280]	"PC.aa.C34.3_T2"	"PC.aa.C34.4_T2"	"PC.aa.C36.0_T2"
## [283]	"PC.aa.C36.1_T2"	"PC.aa.C36.2_T2"	"PC.aa.C36.3_T2"
## [286]	"PC.aa.C36.4_T2"	"PC.aa.C36.5_T2"	"PC.aa.C38.0_T2"
## [289]	"PC.aa.C38.1_T2"	"PC.aa.C38.3_T2"	"PC.aa.C38.4_T2"
## [292]	"PC.aa.C38.5_T2"	"PC.aa.C38.6_T2"	"PC.aa.C40.1_T2"
## [295]	"PC.aa.C40.2_T2"	"PC.aa.C40.3_T2"	"PC.aa.C40.4_T2"
## [298]	"PC.aa.C40.5_T2"	"PC.aa.C40.6_T2"	"PC.aa.C42.0_T2"
## [301]	"PC.aa.C42.1_T2"	"PC.aa.C42.2_T2"	"PC.aa.C42.4_T2"
## [304]	"PC.aa.C42.5_T2"	"PC.aa.C42.6_T2"	"PC.ae.C30.0_T2"
## [307]	"PC.ae.C32.1_T2"	"PC.ae.C32.2_T2"	"PC.ae.C34.0_T2"
## [310]	"PC.ae.C34.1_T2"	"PC.ae.C34.2_T2"	"PC.ae.C34.3_T2"
## [313]	"PC.ae.C36.0_T2"	"PC.ae.C36.1_T2"	"PC.ae.C36.2_T2"
## [316]	"PC.ae.C36.3_T2"	"PC.ae.C36.4_T2"	"PC.ae.C36.5_T2"
## [319]	"PC.ae.C38.0_T2"	"PC.ae.C38.2_T2"	"PC.ae.C38.3_T2"
## [322]	"PC.ae.C38.4_T2"	"PC.ae.C38.5_T2"	"PC.ae.C38.6_T2"
## [325]	"PC.ae.C40.1_T2"	"PC.ae.C40.2_T2"	"PC.ae.C40.3_T2"
## [328]	"PC.ae.C40.4_T2"	"PC.ae.C40.5_T2"	"PC.ae.C40.6_T2"
## [331]	"PC.ae.C42.1_T2"	"PC.ae.C42.2_T2"	"PC.ae.C42.3_T2"

## [334]	"PC.ae.C42.4_T2"	"PC.ae.C42.5_T2"	"PC.ae.C44.3_T2"
## [337]	"PC.ae.C44.4_T2"	"PC.ae.C44.5_T2"	"PC.ae.C44.6_T2"
## [340]	"SM.OH.C14.1_T2"	"SM.OH.C16.1_T2"	"SM.OH.C22.1_T2"
## [343]	"SM.OH.C22.2_T2"	"SM.OH.C24.1_T2"	"SM.C16.0_T2"
## [346]	"SM.C16.1_T2"	"SM.C18.0_T2"	"SM.C18.1_T2"
## [349]	"SM.C20.2_T2"	"SM.C24.0_T2"	"SM.C24.1_T2"
## [352]	"MEDDM_T4"	"MEDCOL_T4"	"MEDINF_T4"
## [355]	"MEDHTA_T4"	"GLU_T4"	"INS_T4"
## [358]	"HOMA_T4"	"HBA1C_T4"	"HBA1C.mmol.mol_T4"
## [361]	"PESO_T4"	"bmi_T4"	"CC_T4"
## [364]	"CINT_T4"	"CAD_T4"	"TAD_T4"
## [367]	"TAS_T4"	"TG_T4"	"COL_T4"
## [370]	"LDL_T4"	"HDL_T4"	"VLDL_T4"
## [373]	"PCR_T4"	"LEP_T4"	"ADIPO_T4"
## [376]	"GOT_T4"	"GPT_T4"	"GGT_T4"
## [379]	"URICO_T4"	"CREAT_T4"	"UREA_T4"
## [382]	"HIERRO_T4"	"TRANSF_T4"	"FERR_T4"
## [385]	"Ile_T4"	"Leu_T4"	"Val_T4"
## [388]	"Ala_T4"	"Pro_T4"	"Gly_T4"
## [391]	"Ser_T4"	"Trp_T4"	"Phe_T4"
## [394]	"Met_T4"	"Orn_T4"	"Arg_T4"
## [397]	"His_T4"	"Asn_T4"	"Asp_T4"
## [400]	"Glu_T4"	"Gln_T4"	"Cit_T4"
## [403]	"Tyr_T4"	"Thr_T4"	"Lys_T4"
## [406]	"Creatinine_T4"	"Kynurenine_T4"	"Putrescine_T4"
## [409]	"Sarcosine_T4"	"Serotonin_T4"	"Taurine_T4"
## [412]	"SDMA_T4"	"CO_T4"	"C2_T4"
## [415]	"C3.OH_T4"	"C6.C4.1.DC.T4"	"C5.DC.C6.OH.T4"
## [418]	"C7.DC_T4"	"C8_T4"	"C10_T4"
## [421]	"C10.1_T4"	"C10.2_T4"	"C14.1_T4"
## [424]	"C14.2_T4"	"C16.1_T4"	"C16.2_T4"
## [427]	"C16.2.OH_T4"	"C18.1_T4"	"C18.1.OH_T4"
## [430]	"C18.2_T4"	"lysoPC.a.C16.0_T4"	"lysoPC.a.C16.1_T4"
## [433]	"lysoPC.a.C17.0_T4"	"lysoPC.a.C18.0_T4"	"lysoPC.a.C18.1_T4"
## [436]	"lysoPC.a.C18.2_T4"	"lysoPC.a.C20.3_T4"	"lysoPC.a.C20.4_T4"
## [439]	"lysoPC.a.C24.0_T4"	"lysoPC.a.C26.0_T4"	"lysoPC.a.C26.1_T4"
## [442]	"lysoPC.a.C28.0_T4"	"lysoPC.a.C28.1_T4"	"PC.aa.C24.0_T4"
## [445]	"PC.aa.C28.1_T4"	"PC.aa.C30.0_T4"	"PC.aa.C32.0_T4"
## [448]	"PC.aa.C32.1_T4"	"PC.aa.C32.3_T4"	"PC.aa.C34.1_T4"
## [451]	"PC.aa.C34.2_T4"	"PC.aa.C34.3_T4"	"PC.aa.C34.4_T4"
## [454]	"PC.aa.C36.0_T4"	"PC.aa.C36.1_T4"	"PC.aa.C36.2_T4"
## [457]	"PC.aa.C36.3_T4"	"PC.aa.C36.4_T4"	"PC.aa.C36.5_T4"
## [460]	"PC.aa.C38.0_T4"	"PC.aa.C38.1_T4"	"PC.aa.C38.3_T4"
## [463]	"PC.aa.C38.4_T4"	"PC.aa.C38.5_T4"	"PC.aa.C38.6_T4"
## [466]	"PC.aa.C40.1_T4"	"PC.aa.C40.2_T4"	"PC.aa.C40.3_T4"
## [469]	"PC.aa.C40.4_T4"	"PC.aa.C40.5_T4"	"PC.aa.C40.6_T4"
## [472]	"PC.aa.C42.0_T4"	"PC.aa.C42.1_T4"	"PC.aa.C42.2_T4"
## [475]	"PC.aa.C42.4_T4"	"PC.aa.C42.5_T4"	"PC.aa.C42.6_T4"
## [478]	"PC.ae.C30.0_T4"	"PC.ae.C32.1_T4"	"PC.ae.C32.2_T4"
## [481]	"PC.ae.C34.0_T4"	"PC.ae.C34.1_T4"	"PC.ae.C34.2_T4"
## [484]	"PC.ae.C34.3_T4"	"PC.ae.C36.0_T4"	"PC.ae.C36.1_T4"
## [487]	"PC.ae.C36.2_T4"	"PC.ae.C36.3_T4"	"PC.ae.C36.4_T4"
## [490]	"PC.ae.C36.5_T4"	"PC.ae.C38.0_T4"	"PC.ae.C38.2_T4"
## [493]	"PC.ae.C38.3_T4"	"PC.ae.C38.4_T4"	"PC.ae.C38.5_T4"

## [496]	"PC.ae.C38.6_T4"	"PC.ae.C40.1_T4"	"PC.ae.C40.2_T4"
## [499]	"PC.ae.C40.3_T4"	"PC.ae.C40.4_T4"	"PC.ae.C40.5_T4"
## [502]	"PC.ae.C40.6_T4"	"PC.ae.C42.1_T4"	"PC.ae.C42.2_T4"
## [505]	"PC.ae.C42.3_T4"	"PC.ae.C42.4_T4"	"PC.ae.C42.5_T4"
## [508]	"PC.ae.C44.3_T4"	"PC.ae.C44.4_T4"	"PC.ae.C44.5_T4"
## [511]	"PC.ae.C44.6_T4"	"SM..OH..C14.1_T4"	"SM..OH..C16.1_T4"
## [514]	"SM..OH..C22.1_T4"	"SM..OH..C22.2_T4"	"SM..OH..C24.1_T4"
## [517]	"SM.C16.0_T4"	"SM.C16.1_T4"	"SM.C18.0_T4"
## [520]	"SM.C18.1_T4"	"SM.C20.2_T4"	"SM.C24.0_T4"
## [523]	"SM.C24.1_T4"	"MEDDM_T5"	"MEDCOL_T5"
## [526]	"MEDINF_T5"	"MEDHTA_T5"	"GLU_T5"
## [529]	"INS_T5"	"HOMA_T5"	"HBA1C_T5"
## [532]	"HBA1C.mmol.mol_T5"	"PESO_T5"	"bmi_T5"
## [535]	"CC_T5"	"CINT_T5"	"CAD_T5"
## [538]	"TAD_T5"	"TAS_T5"	"TG_T5"
## [541]	"COL_T5"	"LDL_T5"	"HDL_T5"
## [544]	"VLDL_T5"	"PCR_T5"	"LEP_T5"
## [547]	"ADIPO_T5"	"GOT_T5"	"GPT_T5"
## [550]	"GGT_T5"	"URICO_T5"	"CREAT_T5"
## [553]	"UREA_T5"	"HIERRO_T5"	"TRANSF_T5"
## [556]	"FERR_T5"	"Ile_T5"	"Leu_T5"
## [559]	"Val_T5"	"Ala_T5"	"Pro_T5"
## [562]	"Gly_T5"	"Ser_T5"	"Trp_T5"
## [565]	"Phe_T5"	"Met_T5"	"Orn_T5"
## [568]	"Arg_T5"	"His_T5"	"Asn_T5"
## [571]	"Asp_T5"	"Glu_T5"	"Gln_T5"
## [574]	"Cit_T5"	"Tyr_T5"	"Thr_T5"
## [577]	"Lys_T5"	"Creatinine_T5"	"Kynurenine_T5"
## [580]	"Putrescine_T5"	"Sarcosine_T5"	"Serotonin_T5"
## [583]	"Taurine_T5"	"SDMA_T5"	"C0_T5"
## [586]	"C2_T5"	"C3.OH_T5"	"C6..C4.1.DC..T5"
## [589]	"C5.DC..C6.OH..T5"	"C7.DC_T5"	"C8_T5"
## [592]	"C10_T5"	"C10.1_T5"	"C10.2_T5"
## [595]	"C14.1_T5"	"C14.2_T5"	"C16.1_T5"
## [598]	"C16.2_T5"	"C16.2.OH_T5"	"C18.1_T5"
## [601]	"C18.1.OH_T5"	"C18.2_T5"	"lysoPC.a.C16.0_T5"
## [604]	"lysoPC.a.C16.1_T5"	"lysoPC.a.C17.0_T5"	"lysoPC.a.C18.0_T5"
## [607]	"lysoPC.a.C18.1_T5"	"lysoPC.a.C18.2_T5"	"lysoPC.a.C20.3_T5"
## [610]	"lysoPC.a.C20.4_T5"	"lysoPC.a.C24.0_T5"	"lysoPC.a.C26.0_T5"
## [613]	"lysoPC.a.C26.1_T5"	"lysoPC.a.C28.0_T5"	"lysoPC.a.C28.1_T5"
## [616]	"PC.aa.C24.0_T5"	"PC.aa.C28.1_T5"	"PC.aa.C30.0_T5"
## [619]	"PC.aa.C32.0_T5"	"PC.aa.C32.1_T5"	"PC.aa.C32.3_T5"
## [622]	"PC.aa.C34.1_T5"	"PC.aa.C34.2_T5"	"PC.aa.C34.3_T5"
## [625]	"PC.aa.C34.4_T5"	"PC.aa.C36.0_T5"	"PC.aa.C36.1_T5"
## [628]	"PC.aa.C36.2_T5"	"PC.aa.C36.3_T5"	"PC.aa.C36.4_T5"
## [631]	"PC.aa.C36.5_T5"	"PC.aa.C38.0_T5"	"PC.aa.C38.1_T5"
## [634]	"PC.aa.C38.3_T5"	"PC.aa.C38.4_T5"	"PC.aa.C38.5_T5"
## [637]	"PC.aa.C38.6_T5"	"PC.aa.C40.1_T5"	"PC.aa.C40.2_T5"
## [640]	"PC.aa.C40.3_T5"	"PC.aa.C40.4_T5"	"PC.aa.C40.5_T5"
## [643]	"PC.aa.C40.6_T5"	"PC.aa.C42.0_T5"	"PC.aa.C42.1_T5"
## [646]	"PC.aa.C42.2_T5"	"PC.aa.C42.4_T5"	"PC.aa.C42.5_T5"
## [649]	"PC.aa.C42.6_T5"	"PC.aa.C30.0_T5"	"PC.aa.C32.1_T5"
## [652]	"PC.aa.C32.2_T5"	"PC.aa.C34.0_T5"	"PC.aa.C34.1_T5"
## [655]	"PC.aa.C34.2_T5"	"PC.aa.C34.3_T5"	"PC.aa.C36.0_T5"


```
## [658] "PC.ae.C36.1_T5"      "PC.ae.C36.2_T5"      "PC.ae.C36.3_T5"
## [661] "PC.ae.C36.4_T5"      "PC.ae.C36.5_T5"      "PC.ae.C38.0_T5"
## [664] "PC.ae.C38.2_T5"      "PC.ae.C38.3_T5"      "PC.ae.C38.4_T5"
## [667] "PC.ae.C38.5_T5"      "PC.ae.C38.6_T5"      "PC.ae.C40.1_T5"
## [670] "PC.ae.C40.2_T5"      "PC.ae.C40.3_T5"      "PC.ae.C40.4_T5"
## [673] "PC.ae.C40.5_T5"      "PC.ae.C40.6_T5"      "PC.ae.C42.1_T5"
## [676] "PC.ae.C42.2_T5"      "PC.ae.C42.3_T5"      "PC.ae.C42.4_T5"
## [679] "PC.ae.C42.5_T5"      "PC.ae.C44.3_T5"      "PC.ae.C44.4_T5"
## [682] "PC.ae.C44.5_T5"      "PC.ae.C44.6_T5"      "SM..OH..C14.1_T5"
## [685] "SM..OH..C16.1_T5"    "SM..OH..C22.1_T5"    "SM..OH..C22.2_T5"
## [688] "SM..OH..C24.1_T5"    "SM.C16.0_T5"         "SM.C16.1_T5"
## [691] "SM.C18.0_T5"         "SM.C18.1_T5"         "SM.C20.2_T5"
## [694] "SM.C24.0_T5"         "SM.C24.1_T5"
```

```
# I ara mirem les 10 primeres columnes i les 5 primeres files d'assays
assay(seMA)[1:5, 1:10]
```

```
##   SUBJECTS SURGERY AGE GENDER Group MEDDM_TO MEDCOL_TO MEDINF_TO MEDHTA_TO
## 1         1 by pass  27      F      1         0         0         0         1
## 2         2 by pass  19      F      2         0         0         0         0
## 3         3 by pass  42      F      1         0         0         0         0
## 4         4 by pass  37      F      2         0         0         0         0
## 5         5 tubular  42      F      1         0         0         0         0
##   GLU_TO
## 1      85
## 2      78
## 3      75
## 4      71
## 5      82
```

```
# I també mirem com ha quedat colData
colData(seMA)
```

```
## DataFrame with 695 rows and 4 columns
##           X      VarName      varType Description
##           <character> <character> <character> <character>
## SUBJECTS      SUBJECTS      SUBJECTS      integer      dataDesc
## SURGERY        SURGERY        SURGERY      character      dataDesc
## AGE            AGE            AGE          integer      dataDesc
## GENDER          GENDER          GENDER      character      dataDesc
## Group          Group          Group        integer      dataDesc
## ...            ...            ...          ...          ...
## SM.C18.0_T5 SM.C18.0_T5 SM.C18.0_T5      numeric      dataDesc
## SM.C18.1_T5 SM.C18.1_T5 SM.C18.1_T5      numeric      dataDesc
## SM.C20.2_T5 SM.C20.2_T5 SM.C20.2_T5      numeric      dataDesc
## SM.C24.0_T5 SM.C24.0_T5 SM.C24.0_T5      numeric      dataDesc
## SM.C24.1_T5 SM.C24.1_T5 SM.C24.1_T5      numeric      dataDesc
```

5. Anàlisi exploratoria de les dades

Com que el joc de dades que he triat conté 395 camps i es fa molt difícil fer un anàlisi complet, provem a fer una cosa una mica més ajustada i agafaré només les dades de descripció dels casos analitzats i alguns metabolits. En concret em pararé en el ferro. És a dir que agafaré les primeres 40 columnes. En un context real s'hauria de fer amb tot i l'anàlisi seria molt llarg i tediós.

Per tant, comencem per fer un subconjunt del nostre SummarizedExperiment i el guardem com un data frame per

poder treballar amb ell.

```
analisi <- seMA[,1:40]
analisi
```

```
## class: SummarizedExperiment
## dim: 39 40
## metadata(0):
## assays(1): counts
## rownames(39): 1 2 ... 38 39
## rowData names(0):
## colnames(40): SUBJECTS SURGERY ... Ile_T0 Leu_T0
## colData names(4): X VarName varTpe Description
```

Ara, per tant, tenim un segon SummarizedExperiment que és un subset del primer. Conté només les dades d'identificació de tots els pacients (columnes 1-9) i els primers 31 metabolits que es van recollir

5.1. Estadístics bàsics

Amb aixó, podem mirar una mica el que tenim

```
str(assay(analisi))
```

```
## 'data.frame':   39 obs. of  40 variables:
## $ SUBJECTS      : int  1 2 3 4 5 6 7 8 9 10 ...
## $ SURGERY       : chr  "by pass" "by pass" "by pass" "by pass" ...
## $ AGE           : int  27 19 42 37 42 24 33 55 40 47 ...
## $ GENDER        : chr  "F" "F" "F" "F" ...
## $ Group         : int  1 2 1 2 1 2 1 1 1 1 ...
## $ MEDDM_T0      : int  0 0 0 0 0 0 0 0 0 0 ...
## $ MEDCOL_T0     : int  0 0 0 0 0 0 0 0 0 0 ...
## $ MEDINF_T0     : int  0 0 0 0 0 0 0 1 0 0 ...
## $ MEDHTA_T0     : int  1 0 0 0 0 0 0 0 0 0 ...
## $ GLU_T0        : int  85 78 75 71 82 71 80 90 92 84 ...
## $ INS_T0        : num  11.4 12.1 8.41 12.8 6.01 9.88 9.2 3.4 5.43 6.98 ...
## $ HOMA_T0       : num  2.4 2.32 1.56 2.25 1.22 1.73 1.82 0.76 1.23 1.45 ...
## $ HBA1C_T0      : num  NA NA 5.4 5.1 5.6 5.1 5.6 5.5 5.7 5.5 ...
## $ HBA1C.mmol.mol_T0: num  NA NA 35.5 32.2 37.7 ...
## $ PESO_T0       : int  151 139 84 136 121 148 109 109 114 120 ...
## $ bmi_T0        : num  62.9 47 29.8 53.1 46.6 48.8 43.7 41.8 44 40.6 ...
## $ CC_T0         : num  0.7 NA 0.7 1 0.9 0.7 0.9 0.9 0.9 NA ...
## $ CINT_T0       : int  116 NA 90 157 123 110 122 124 136 NA ...
## $ CAD_T0        : int  167 NA 126 162 132 148 141 136 148 NA ...
## $ TAD_T0        : int  125 NA 79 73 84 74 65 75 91 NA ...
## $ TAS_T0        : int  174 NA 111 127 122 131 121 128 128 NA ...
## $ TG_T0         : int  147 150 45 109 30 61 75 53 164 145 ...
## $ COL_T0        : int  256 180 211 205 102 121 192 181 154 220 ...
## $ LDL_T0        : num  167 94 114 146 24 60.8 96 99.4 67.2 160 ...
## $ HDL_T0        : int  60 56 88 37 72 48 81 71 54 31 ...
## $ VLDL_T0       : num  29.4 30 9 21.8 6 12.2 15 10.6 32.8 29 ...
## $ PCR_T0        : num  10.2 9 3.05 8.89 NA 1.6 4.52 NA NA NA ...
## $ LEP_T0        : num  155 84 27 46 NA 38 61 NA NA NA ...
## $ ADIPO_T0      : num  8.15 7.94 16.7 4.68 NA 11.8 6.51 NA NA NA ...
## $ GOT_T0        : int  21 16 24 21 42 11 14 50 21 15 ...
## $ GPT_T0        : int  33 40 39 37 68 26 25 24 11 30 ...
## $ GGT_T0        : int  22 25 20 20 18 12 16 36 16 52 ...
```

```
## $ URICO_TO      : num  5.7 NA 2.7 4.9 5.2 3.7 3.8 5.6 5.1 5 ...
## $ CREAT_TO      : num  0.8 0.8 0.8 0.7 0.8 0.7 0.6 0.7 0.7 1 ...
## $ UREA_TO       : int   33 29 28 22 29 14 26 49 24 24 ...
## $ HIERRO_TO     : int   77 113 142 64 94 102 72 13 36 113 ...
## $ TRANSF_TO     : int   NA 290 251 289 252 200 NA 300 288 235 ...
## $ FERR_TO       : num  53 25 15 18 39.4 25 24 23 48 101 ...
## $ Ile_TO        : num  53.9 95.7 57.8 84.5 72 109 73.3 91.7 83.8 113 ...
## $ Leu_TO        : num  105 188 119 180 140 205 147 191 183 225 ...
```

```
summary(assay(analisi))
```

```
##      SUBJECTS      SURGERY      AGE      GENDER
## Min.   : 1.0   Length:39      Min.   :19.00   Length:39
## 1st Qu.:10.5   Class :character  1st Qu.:35.00   Class :character
## Median :20.0   Mode  :character  Median :41.00   Mode  :character
## Mean   :20.0
## 3rd Qu.:29.5
## Max.   :39.0
##
##      Group      MEDDM_TO  MEDCOL_TO      MEDINF_TO
## Min.   :1.000   Min.   :0   Min.   :0.00000   Min.   :0.0000
## 1st Qu.:1.000   1st Qu.:0   1st Qu.:0.00000   1st Qu.:0.0000
## Median :1.000   Median :0   Median :0.00000   Median :0.0000
## Mean   :1.385   Mean   :0   Mean   :0.02632   Mean   :0.1316
## 3rd Qu.:2.000   3rd Qu.:0   3rd Qu.:0.00000   3rd Qu.:0.0000
## Max.   :2.000   Max.   :0   Max.   :1.00000   Max.   :1.0000
##
##      NA's      :1   NA's      :1   NA's      :1
##
##      MEDHTA_TO      GLU_TO      INS_TO      HOMA_TO
## Min.   :0.0000   Min.   : 71.0   Min.   : 3.40   Min.   : 0.760
## 1st Qu.:0.0000   1st Qu.: 91.0   1st Qu.:11.85   1st Qu.: 2.435
## Median :0.0000   Median :103.0   Median :16.00   Median : 4.210
## Mean   :0.2368   Mean   :106.5   Mean   :17.60   Mean   : 4.890
## 3rd Qu.:0.0000   3rd Qu.:109.5   3rd Qu.:21.15   3rd Qu.: 5.720
## Max.   :1.0000   Max.   :263.0   Max.   :43.00   Max.   :13.600
##
##      NA's      :1
##
##      HBA1C_TO      HBA1C.mmol.mol_TO      PESO_TO      bmi_TO
## Min.   :5.100   Min.   :32.23   Min.   : 84.0   Min.   :29.80
## 1st Qu.:5.400   1st Qu.:35.51   1st Qu.:119.5   1st Qu.:44.40
## Median :5.600   Median :37.69   Median :135.0   Median :48.80
## Mean   :5.592   Mean   :37.60   Mean   :140.0   Mean   :50.52
## 3rd Qu.:5.800   3rd Qu.:39.88   3rd Qu.:155.0   3rd Qu.:55.35
## Max.   :6.400   Max.   :46.44   Max.   :200.0   Max.   :68.60
##
##      NA's      :15   NA's      :15
##
##      CC_TO      CINT_TO      CAD_TO      TAD_TO
## Min.   :0.7000   Min.   : 90.0   Min.   : 85.0   Min.   : 60.00
## 1st Qu.:0.9000   1st Qu.:122.5   1st Qu.:136.0   1st Qu.: 74.50
## Median :0.9000   Median :133.0   Median :148.0   Median : 83.00
## Mean   :0.9343   Mean   :135.7   Mean   :147.1   Mean   : 83.58
## 3rd Qu.:1.0000   3rd Qu.:147.5   3rd Qu.:158.5   3rd Qu.: 90.00
## Max.   :1.7000   Max.   :186.0   Max.   :182.0   Max.   :125.00
##
##      NA's      :4   NA's      :4   NA's      :4   NA's      :8
##
##      TAS_TO      TG_TO      COL_TO      LDL_TO
## Min.   :111.0   Min.   : 30.0   Min.   :102.0   Min.   : 24.0
## 1st Qu.:121.0   1st Qu.: 84.0   1st Qu.:180.5   1st Qu.:100.3
## Median :128.0   Median :118.0   Median :207.0   Median :130.0
```

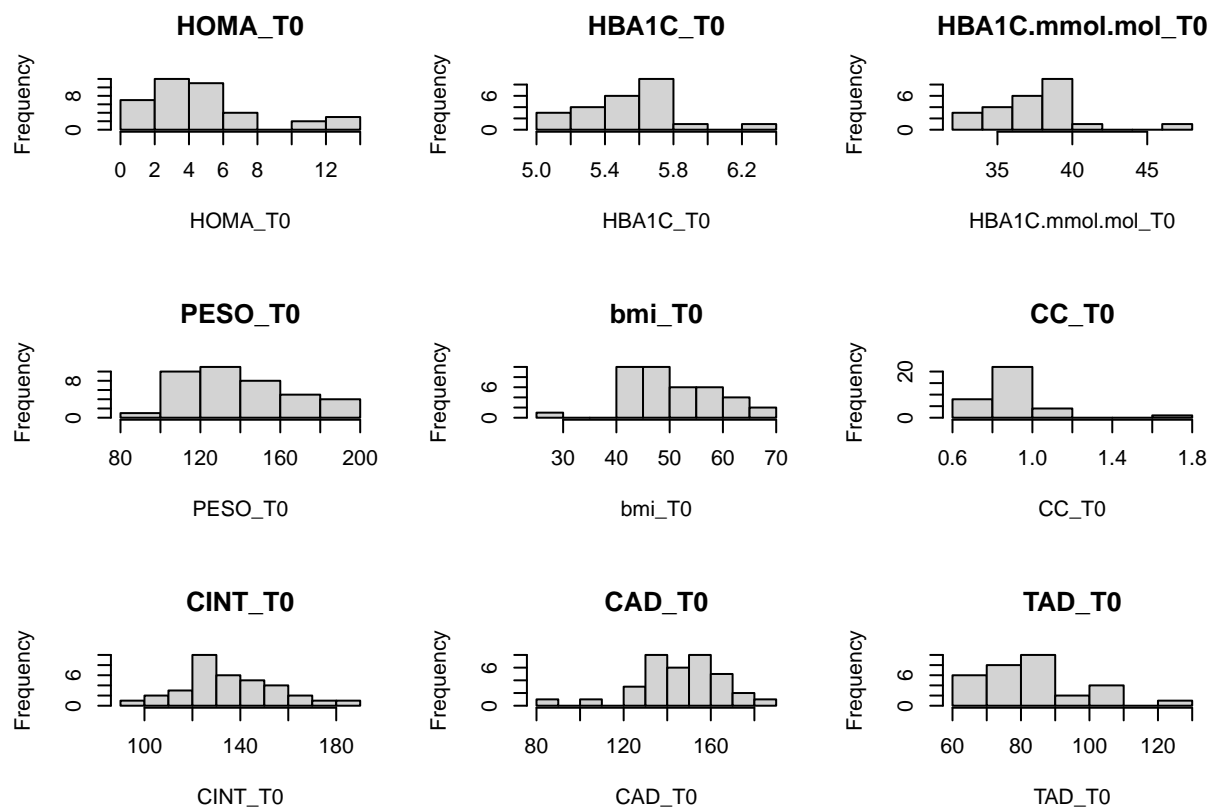
```
## Mean :133.5 Mean :129.1 Mean :206.6 Mean :128.2
## 3rd Qu.:144.0 3rd Qu.:151.0 3rd Qu.:236.0 3rd Qu.:158.8
## Max. :180.0 Max. :268.0 Max. :295.0 Max. :191.0
## NA's :8
## HDL_T0 VLDL_T0 PCR_T0 LEP_T0
## Min. : 11.00 Min. : 6.00 Min. : 0.00 Min. : 27.00
## 1st Qu.: 39.25 1st Qu.:16.80 1st Qu.: 3.51 1st Qu.: 49.25
## Median : 47.00 Median :23.60 Median : 8.25 Median : 67.20
## Mean : 49.97 Mean :25.81 Mean :10.13 Mean : 78.33
## 3rd Qu.: 59.50 3rd Qu.:30.20 3rd Qu.:13.90 3rd Qu.: 98.25
## Max. :107.00 Max. :53.60 Max. :29.80 Max. :155.00
## NA's :1 NA's :14 NA's :13
## ADIPO_T0 GOT_T0 GPT_T0 GGT_T0
## Min. : 1.720 Min. :11.0 Min. : 11.00 Min. :12.00
## 1st Qu.: 4.700 1st Qu.:15.0 1st Qu.: 31.00 1st Qu.:20.00
## Median : 7.460 Median :20.0 Median : 40.00 Median :26.00
## Mean : 7.702 Mean :23.9 Mean : 44.54 Mean :31.41
## 3rd Qu.: 8.590 3rd Qu.:25.0 3rd Qu.: 48.50 3rd Qu.:37.00
## Max. :17.100 Max. :83.0 Max. :157.00 Max. :86.00
## NA's :14 NA's :2
## URICO_T0 CREAT_T0 UREA_T0 HIERRO_T0
## Min. :2.700 Min. :0.5000 Min. :14.00 Min. : 13.00
## 1st Qu.:4.700 1st Qu.:0.7000 1st Qu.:24.00 1st Qu.: 50.00
## Median :5.500 Median :0.8000 Median :29.00 Median : 69.00
## Mean :5.501 Mean :0.7923 Mean :31.38 Mean : 75.61
## 3rd Qu.:6.200 3rd Qu.:0.9000 3rd Qu.:38.00 3rd Qu.: 92.25
## Max. :8.600 Max. :1.2000 Max. :56.00 Max. :166.00
## NA's :2 NA's :1
## TRANSF_T0 FERR_T0 Ile_T0 Leu_T0
## Min. :159.0 Min. : 11.10 Min. : 53.50 Min. : 90.1
## 1st Qu.:236.5 1st Qu.: 25.00 1st Qu.: 83.55 1st Qu.:150.5
## Median :268.5 Median : 49.00 Median : 91.70 Median :183.0
## Mean :262.9 Mean : 77.06 Mean : 98.98 Mean :185.2
## 3rd Qu.:289.8 3rd Qu.: 97.00 3rd Qu.:116.00 3rd Qu.:217.0
## Max. :374.0 Max. :372.00 Max. :150.00 Max. :281.0
## NA's :9 NA's :2
```

Ja només amb aquesta primera mirada, podem comprovar que les dades tenen una enorme variabilitat de rangs el que pot portar algun problema a l'hora de fer anàlisi més adelantats i que calgui escalar les variables en alguns casos

5.2. Anàlisi univariant

Ho fem amb histogrames. Farem 9 de les variables per veure una mica com es distribueixen

```
opt <- par(mfrow=c(3,3))
for (i in 12:20)
  hist(assay(analisi)[,i], xlab = colData(analisi)[i,2], main = colData(analisi)[i,2])
```

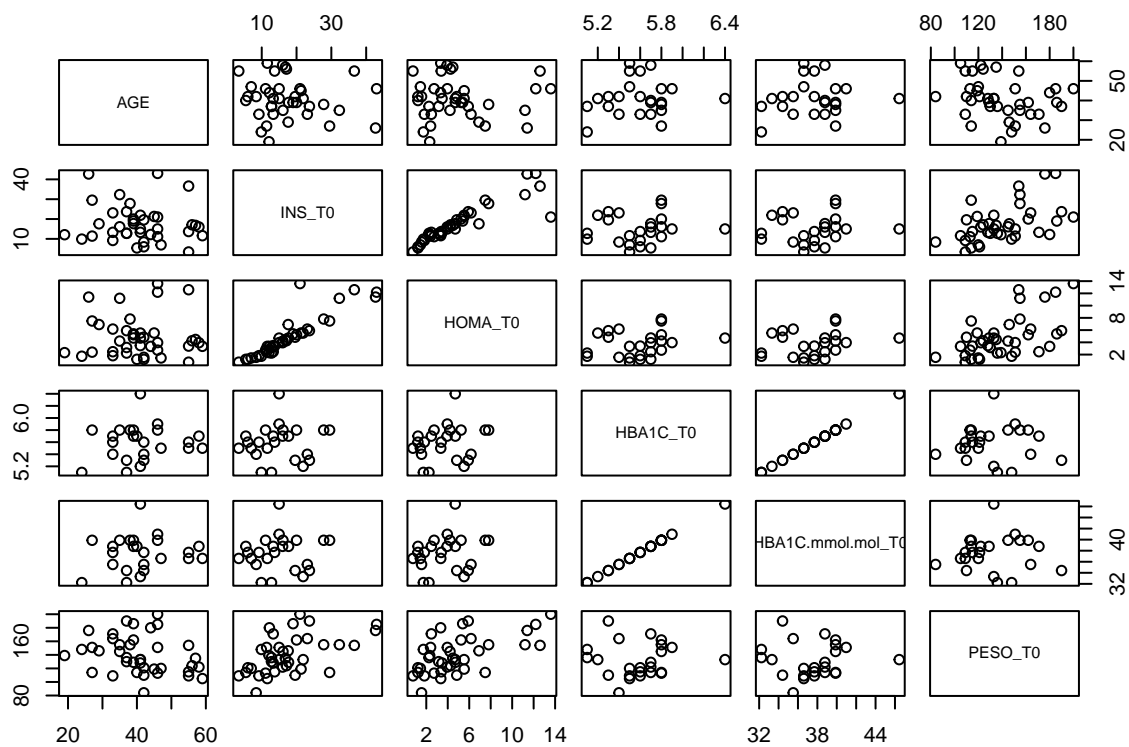


```
par(opt)
```

5.3. Anàlisi de correlacions

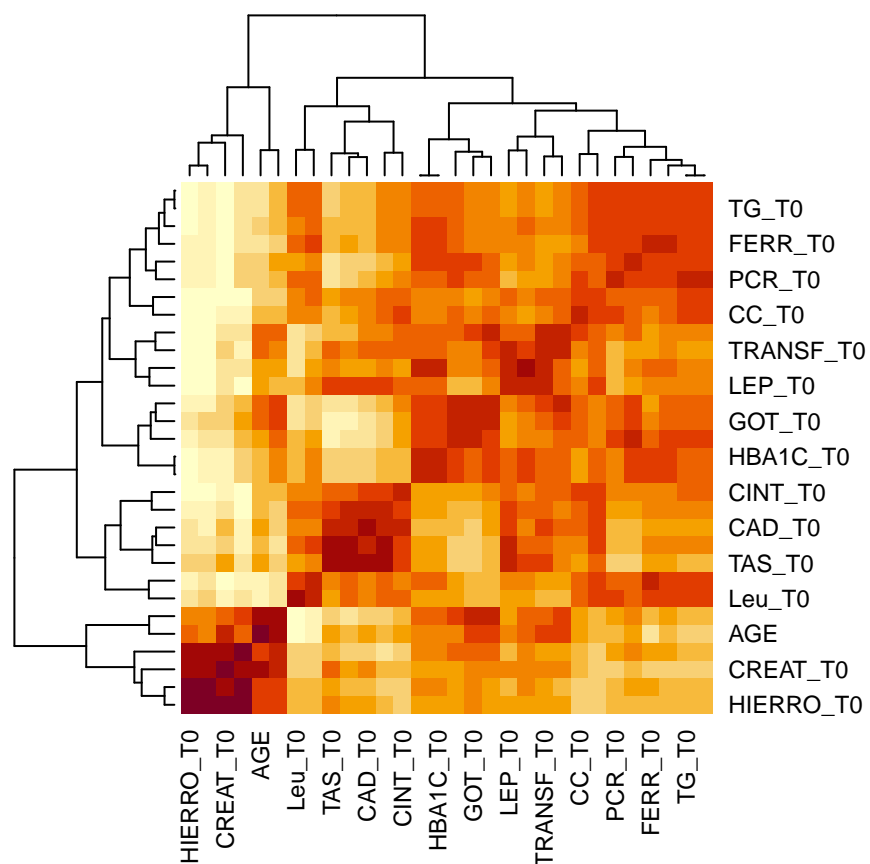
Ho farem amb plot. Un exemple

```
# Mirem per exemple, si l'edat es correlaciona be amb alguns
plot(assay(analisi)[,c(3,11:15)])
```



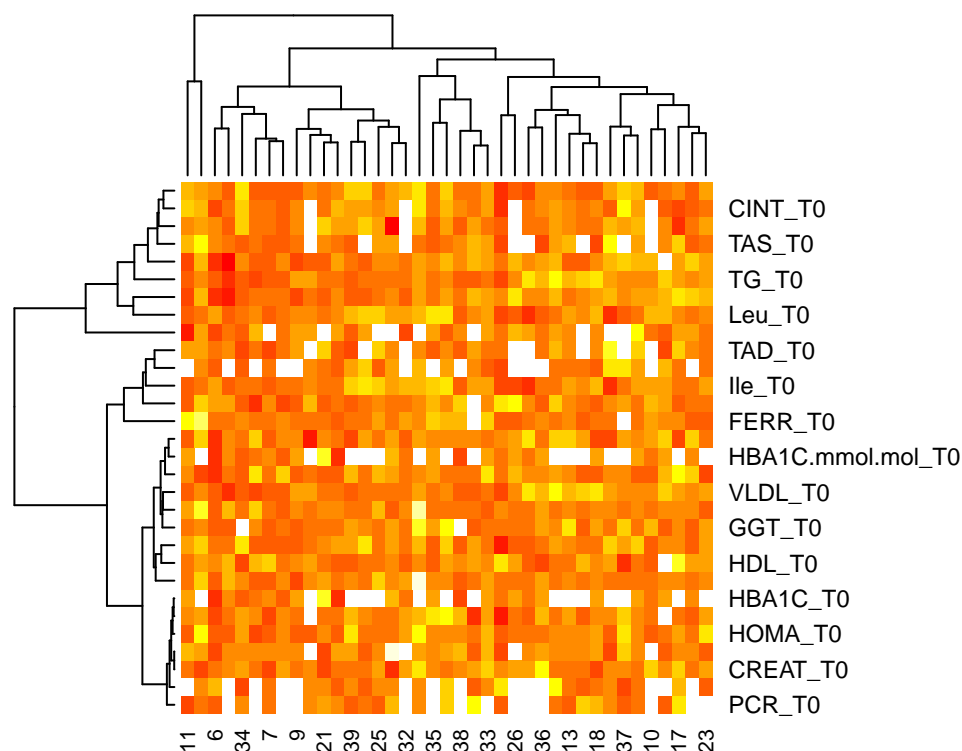
Una altra opció pot ser un heatmap de la matriu de correlacions que permetria veure totes les relacions

```
cor_matrix <- cor(na.omit(assay(analisi)[,c(3,12:40)]))
heatmap(cor_matrix)
```



Un altre heatmap que calcula distancies

```
dist_matrix <- t(assay(analisi)[,c(3,12:40)])
heatmap(dist_matrix, col=heat.colors(16))
```



5.4. Components principals

Primer hem d'eliminar els NA i despres podem. Cal fer-ho amb les columnes de metabolits (11-40)

```
PCAMA <- prcomp(na.omit(assay(analisi)[,c(3,12:40)]), scale. = TRUE)
PCAMA
```

```
## Standard deviations (1, ..., p=5):
## [1] 3.869264e+00 2.708545e+00 2.318885e+00 1.521628e+00 1.443830e-15
##
## Rotation (n x k) = (30 x 5):
##
```

	PC1	PC2	PC3	PC4
## AGE	-3.879932e-02	0.22167989	-0.3296551604	-0.118639705
## HOMA_T0	2.361425e-01	0.07777135	0.1125121813	0.150887864
## HBA1C_T0	1.848513e-01	0.18356771	-0.0107162236	0.322363548
## HBA1C.mmol.mol_T0	1.848247e-01	0.18361534	-0.0107213019	0.322375631
## PESO_T0	1.030985e-01	-0.29665719	-0.1728637006	0.122169797
## bmi_T0	1.771316e-01	-0.22673533	-0.1680474258	0.023419549
## CC_T0	2.207149e-01	0.00758126	-0.0077235123	-0.341444926
## CINT_T0	1.957452e-01	-0.12561526	-0.1496859137	-0.286552989
## CAD_T0	1.112784e-01	-0.23777875	-0.2598588414	-0.125945898
## TAD_T0	1.607530e-01	0.08101446	-0.2027032128	0.385468271
## TAS_T0	6.116069e-02	-0.23731455	-0.2914386581	0.178911570
## TG_T0	2.382890e-01	0.05363323	0.1546591555	-0.008992496
## COL_T0	1.555571e-01	0.26556346	-0.1365157389	-0.093253755
## LDL_T0	1.731153e-01	0.14942926	-0.2412912905	-0.179315548


```

## HDL_TO          -1.731337e-01  0.25966592  0.0425202085  0.142259672
## VLDL_TO         2.382890e-01  0.05363323  0.1546591555 -0.008992496
## PCR_TO          2.012443e-01  0.09770185  0.2063896433 -0.202128531
## LEP_TO          1.695036e-01 -0.06206151 -0.2769408895  0.236209979
## ADIPO_TO        -2.276517e-01  0.11811067  0.0532546355  0.214491905
## GOT_TO          1.227601e-01  0.31588754 -0.0201212506 -0.131704788
## GPT_TO          1.688682e-01  0.26844697  0.0683883809 -0.091181103
## GGT_TO          2.091045e-01  0.18057083  0.1401439169  0.015756329
## URICO_TO        2.537605e-01 -0.06332388 -0.0248458392 -0.037185033
## CREAT_TO        -2.097037e-01  0.10713755 -0.2180727821  0.027140380
## UREA_TO         2.128422e-05  0.32536065 -0.1981896680 -0.072525585
## HIERRO_TO       -2.444839e-01  0.10250032 -0.0003968416  0.110083675
## TRANSF_TO       1.625339e-01  0.07166917 -0.3235396869  0.041264281
## FERR_TO         2.149934e-01  0.01261204  0.1882238050  0.224145251
## Ile_TO          1.941765e-01 -0.15392289  0.1743106342  0.206072925
## Leu_TO          1.372623e-01 -0.20988123  0.2701868494 -0.030995686
##                  PC5
## AGE              0.225440592
## HOMA_TO          -0.144025054
## HBA1C_TO         0.462598015
## HBA1C.mmol.mol_TO 0.008170488
## PESO_TO         -0.219322952
## bmi_TO           0.164633909
## CC_TO            0.098363609
## CINT_TO          0.057801368
## CAD_TO           -0.112706509
## TAD_TO           0.212400370
## TAS_TO           0.041880828
## TG_TO            -0.058572235
## COL_TO           0.126291305
## LDL_TO           -0.132613519
## HDL_TO           -0.296539028
## VLDL_TO          -0.032781106
## PCR_TO           0.043376648
## LEP_TO           -0.203457228
## ADIPO_TO         -0.146210377
## GOT_TO           -0.165346454
## GPT_TO           0.232846351
## GGT_TO           -0.186318562
## URICO_TO         -0.163492538
## CREAT_TO         0.228432155
## UREA_TO          -0.183839340
## HIERRO_TO        -0.090895613
## TRANSF_TO        -0.281021101
## FERR_TO          -0.196905482
## Ile_TO           0.135385332
## Leu_TO           0.051283737

```

summary(PCAMA)

```

## Importance of components:
##                  PC1    PC2    PC3    PC4    PC5
## Standard deviation 3.869 2.7085 2.3189 1.52163 1.444e-15
## Proportion of Variance 0.499 0.2445 0.1792 0.07718 0.000e+00
## Cumulative Proportion 0.499 0.7436 0.9228 1.00000 1.000e+00

```

Les deux premières composantes expliquent 74% de la variance et les 3 premières, 93%. Graphiquons les deux premières.

```
loads <- round(PCAMA$sdev^2/sum(PCAMA$sdev^2)*100,1)
groupColors <- c(rep("red", 4), rep("blue", 4))
xlab <- c(paste("PC1",loads[1],"%"))
ylab <- c(paste("PC2",loads[2],"%"))
plot(PCAMA$x[,1:2],
     xlab=xlab,
     ylab=ylab,
     col=groupColors,
     main = "Principal components (PCA)")
filesSenseNA <- complete.cases(assay(analisi)[,c(3,12:40)])
names2plot <- which(filesSenseNA)

text(PCAMA$x[,1],PCAMA$x[,2],names2plot, pos=3, cex=.6)
```

