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Informe PEC1 Análisis Datos Ómicos

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1- Descarga de los datos

Descargamos los datos de human cachexia del repositorio de github y los importamos al proyecto.

```
human_cachexia<-read.csv("human_cachexia.csv")
```

Ejecutamos class para saber de que forma esta organizada la información

```
class(human_cachexia)
```

```
## [1] "data.frame"
```

2- Abstract

Human_cachexia de metaboAnalyst se trata de una tabla de concentración de metabolitos en muestras de orina humana de dos grupos (cachexic y control).

La caquexia se trata de un síndrome metabólico complicado y relacionado con una enfermedad subyacente caracterizado por la perdida de masa muscular (con o sin perdida de masa grasa).

El dataframe human_cachexia probablemente sea el resultado de un análisis metabólico comparativo entre pacientes con caquexia y un grupo control (quizá en un intento de encontrar metabolitos en orina humana que puedan servir como un indicador del síndrome).

3- Objetivo del estudio

Teniendo en cuenta que no tenemos información del objetivo puesto que carecemos de metadatos que lo describan u otras consideraciones podemos realizar una inferencia del objetivo final del estudio:

¿Hay diferencias significativas para la cantidad de alguno del listado de metabolitos detectados en orina entre los grupos Control y Cachexia?

4- Exploración de los datos

Realizamos str para hacernos una idea general de la estructura interna del data frame.

```
str(human_cachexia)
```

```

## 'data.frame':    77 obs. of  65 variables:
## $ Patient.ID          : chr  "PIF_178" "PIF_087" "PIF_090" "NETL_005_V1" ...
## $ Muscle.loss          : chr  "cachexic" "cachexic" "cachexic" "cachexic" ...
## $ X1.6.Anhydro.beta.D.glucose: num  40.9 62.2 270.4 154.5 22.2 ...
## $ X1.Methylnicotinamide: num  65.4 340.4 64.7 53 73.7 ...
## $ X2.Aminobutyrate     : num  18.7 24.3 12.2 172.4 15.6 ...
## $ X2.Hydroxyisobutyrate: num  26.1 41.7 65.4 74.4 83.9 ...
## $ X2.Oxoglutarate      : num  71.5 67.4 23.8 1199.9 33.1 ...
## $ X3.Aminoisobutyrate  : num  1480.3 116.8 14.3 555.6 29.7 ...
## $ X3.Hydroxybutyrate   : num  56.83 43.82 5.64 175.91 76.71 ...
## $ X3.Hydroxyisovalerate: num  10.1 79.8 23.3 25 69.4 ...
## $ X3.Indoxylsulfate    : num  567 369 665 412 166 ...
## $ X4.Hydroxyphenylacetate: num  120.3 432.7 292.9 214.9 97.5 ...
## $ Acetate              : num  126.5 212.7 314.2 37.3 407.5 ...
## $ Acetone              : num  9.49 11.82 4.44 206.44 44.26 ...
## $ Adipate              : num  38.1 327 131.6 144 15 ...
## $ Alanine              : num  314 871 464 590 1119 ...
## $ Asparagine            : num  159.2 157.6 89.1 273.1 42.5 ...
## $ Betaine              : num  110 245 117 279 392 ...
## $ Carnitine             : num  265.1 120.3 25 200.3 84.8 ...
## $ Citrate               : num  3714 2618 863 13630 854 ...
## $ Creatine              : num  196.4 212.7 221.4 85.6 105.6 ...
## $ Creatinine             : num  16482 15835 24588 20952 6768 ...
## $ Dimethylamine          : num  633 608 735 1064 242 ...
## $ Ethanolamine           : num  645 488 407 821 365 ...
## $ Formate               : num  441 252 250 469 114 ...
## $ Fucose                : num  337 198.3 186.8 407.5 26.1 ...
## $ Fumarate              : num  7.69 18.92 7.1 96.54 19.69 ...
## $ Glucose               : num  395 8691 1353 863 6836 ...
## $ Glutamine              : num  871 602 302 1686 433 ...
## $ Glycine               : num  2039 1108 620 5064 395 ...
## $ Glycolate              : num  685.4 652 141.2 70.8 26.6 ...
## $ Guanidoacetate         : num  154 110 183 103 53 ...
## $ Hippurate              : num  4582 1737 4316 757 1153 ...
## $ Histidine              : num  925 846 284 1043 327 ...
## $ Hypoxanthine            : num  97.5 82.3 114.4 223.6 66.7 ...
## $ Isoleucine              : num  5.58 8.17 9.3 37.71 40.04 ...
## $ Lactate                : num  107 369 750 369 3641 ...
## $ Leucine                 : num  42.1 77.5 31.5 103.5 101.5 ...
## $ Lysine                  : num  146.9 284.3 97.5 290 122.7 ...
## $ Methylamine              : num  52.5 23.6 18.7 48.9 27.9 ...
## $ Methylguanidine         : num  9.97 7.69 4.66 141.17 5.31 ...
## $ N.N.Dimethylglycine    : num  23.3 87.4 24.5 40 46.1 ...
## $ O.Acetylcarnitine       : num  52.98 50.4 5.58 254.68 45.6 ...
## $ Pantothenate            : num  25.8 186.8 145.5 42.5 74.4 ...
## $ Pyroglutamate           : num  437 437 713 567 185 ...
## $ Pyruvate                : num  21.1 37 29.4 64.1 12.3 ...
## $ Quinolinate             : num  165.7 73 192.5 86.5 38.1 ...
## $ Serine                  : num  284 392 296 1249 206 ...
## $ Succinate               : num  154.5 244.7 142.6 144 68.7 ...
## $ Sucrose                  : num  45.1 459.4 160.8 111 75.2 ...
## $ Tartrate                 : num  97.51 32.79 16.28 837.15 4.53 ...
## $ Taurine                  : num  1920 1261 4273 1525 469 ...
## $ Threonine                : num  184.9 198.3 110 376.1 64.1 ...
## $ Trigonelline             : num  943.9 208.5 192.5 992.3 86.5 ...

```

```
## $ Trimethylamine.N.oxide      : num 2122 639 1153 1451 172 ...
## $ Tryptophan                 : num 259.8 83.1 82.3 235.1 103.5 ...
## $ Tyrosine                   : num 290 167.3 60.3 323.8 142.6 ...
## $ Uracil                      : num 111 47 31.5 30.6 44.3 ...
## $ Valine                      : num 86.5 110 59.1 102.5 160.8 ...
## $ Xylose                      : num 72.2 192.5 2164.6 125.2 186.8 ...
## $ cis.Aconitate               : num 237 334 330 1863 101 ...
## $ myo.Inositol                : num 135.6 376.1 86.5 247.2 750 ...
## $ trans.Aconitate              : num 51.9 217 58.6 75.9 98.5 ...
## $ pi.Methylhistidine          : num 157.6 308 145.5 249.6 84.8 ...
## $ tau.Methylhistidine         : num 160.8 130.3 83.9 254.7 79.8 ...
```

Observamos que se trata de un dataframe con 65 columnas, 63 de ellas son numéricas y dos de "characters"

4.1- SummarizedExperiment object

Para generar el objeto SummarizedExperiment debemos importar la librería de SummarizedExperiment.

```
if (!requireNamespace("BiocManager", quietly = TRUE)) {
  install.packages("BiocManager")
}
BiocManager::install("SummarizedExperiment")
```

```
## Bioconductor version 3.20 (BiocManager 1.30.25), R 4.4.0 (2024-04-24 ucrt)
```

```
## Warning: package(s) not installed when version(s) same as or greater than current; use
##   `force = TRUE` to re-install: 'SummarizedExperiment'
```

```
## Installation paths not writeable, unable to update packages
##   path: C:/Program Files/R/R-4.4.0/library
##   packages:
##     boot, foreign, KernSmooth, MASS, Matrix, nlme, survival
```

```
## Old packages: 'Biobase', 'GenomicRanges', 'httr2', 'IRanges', 'S4Arrays',
##   'S4Vectors', 'xfun', 'XVector', 'zlibbioc'
```

```
library(SummarizedExperiment)
```

```
## Warning: package 'SummarizedExperiment' was built under R version 4.4.1
```

```
## Cargando paquete requerido: MatrixGenerics
```

```
## Warning: package 'MatrixGenerics' was built under R version 4.4.1
```

```
## Cargando paquete requerido: matrixStats
```

```
## Warning: package 'matrixStats' was built under R version 4.4.1
```

```
##  
## Adjuntando el paquete: 'MatrixGenerics'
```

```
## The following objects are masked from 'package:matrixStats':  
##  
## colAlls, colAnyNAs, colAnys, colAvgsPerRowSet, 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, rowAvgsPerColSet,  
## 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
```

```
## Cargando paquete requerido: GenomicRanges
```

```
## Cargando paquete requerido: stats4
```

```
## Cargando paquete requerido: BiocGenerics
```

```
## Warning: package 'BiocGenerics' was built under R version 4.4.1
```

```
##  
## Adjuntando el paquete: '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
```

```
## Cargando paquete requerido: S4Vectors
```

```
## Warning: package 'S4Vectors' was built under R version 4.4.1
```

```
##  
## Adjuntando el paquete: 'S4Vectors'
```

```
## The following object is masked from 'package:utils':  
##  
##     findMatches
```

```
## The following objects are masked from 'package:base':  
##  
##     expand.grid, I, uname
```

```
## Cargando paquete requerido: IRanges
```

```
## Warning: package 'IRanges' was built under R version 4.4.1
```

```
##  
## Adjuntando el paquete: 'IRanges'
```

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

```
## Cargando paquete requerido: GenomeInfoDb
```

```
## Warning: package 'GenomeInfoDb' was built under R version 4.4.1
```

```
## Cargando paquete requerido: Biobase
```

```
## Welcome to Bioconductor  
##  
##     Vignettes contain introductory material; view with  
##     'browseVignettes()'. To cite Bioconductor, see  
##     'citation("Biobase")', and for packages 'citation("pkgname")'.
```

```
##  
## Adjuntando el paquete: 'Biobase'
```

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

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

```
## Warning: replacing previous import 'S4Arrays::read_block' by
## 'DelayedArray::read_block' when loading 'SummarizedExperiment'
```

Generamos un data frame solo con los datos numéricos de human_cachexia transpuestos, puesto que para poder generar el object summarizedexperiment las variables están en las filas y las samples en las columnas y en nuestro dataframe están al revés.

Con esto tendremos los datos del assay.

Después generamos el dataframe para col_data y la lista de nombres para colnames.

```
datos_cachexiat <- t(as.matrix(human_cachexia[, -c(1, 2)]))
colnames(datos_cachexiat) <- human_cachexia$Patient.ID
col_data <- DataFrame(Patient.ID = human_cachexia$Patient.ID,
                      Muscle.loss = human_cachexia$Muscle.loss)

se <- SummarizedExperiment(
  assays = list(counts = datos_cachexiat),
  colData = col_data
)
se
```

```
## class: SummarizedExperiment
## dim: 63 77
## metadata(0):
## assays(1): counts
## rownames(63): X1.6.Anhydro.beta.D.glucose X1.Methylnicotinamide ...
##   pi.Methylhistidine tau.Methylhistidine
## rowData names(0):
## colnames(77): PIF_178 PIF_087 ... NETL_003_V1 NETL_003_V2
## colData names(2): Patient.ID Muscle.loss
```

Por último, creamos el objeto en formato binario para añadirlo al repositorio de github:

```
save(se, file = "summarized_experiment.Rda")
```

4.2- Análisis estadístico

Para realizar el análisis estadístico que responda a la pregunta biológica planteada en el objetivo del estudio realizaremos un conjunto de estudios.

En los datos proporcionados en el repositorio de github nos indican que se tratan de datos desapareados (diferente proporción de control y cachexic). Lo comprobamos mediante:

```
table(human_cachexia$Muscle.loss)
```

```
##
## cachexic  control
##      47       30
```

Tendremos que tener este dato en cuenta a la hora de realizar una análisis multivariante. Empezamos por crear un data.frame solo con las variables numéricas, sin transponerlas.

```
datos_cachexia<-human_cachexia[, -c(1:2)]
```

Después pasamos la variable Muscle.loss de un character a factor para poder utilizarlo en las funciones.
Realizamos un estudio general de la media, la mean y la desviación estándar.

La revisión de estos datos ya nos puede dar una idea general de las diferencias para los metabolitos entre los dos grupos.

```
muscle<-as.factor(human_cachexia$Muscle.loss)
apply(datos_cachexia, MARGIN = 2, FUN = tapply, INDEX=muscle, mean)
```

```
##          X1.6.Anhydro.beta.D.glucose X1.Methylnicotinamide X2.Aminobutyrate
## cachexic           128.68894          70.56426       23.669149
## control            69.50533          73.15500        9.528333
##          X2.Hydroxyisobutyrate X2.Oxoglutarate X3.Aminoisobutyrate
## cachexic           43.23766          183.11043       100.2747
## control            27.87100          85.51733        39.9110
##          X3.Hydroxybutyrate X3.Hydroxyisovalerate X3.Indoxylsulfate
## cachexic           29.260638          27.60638       265.1577
## control             9.898667          12.31267       146.3763
##          X4.Hydroxyphenylacetate Acetate Acetone Adipate Alanine
## cachexic           119.82255          85.63298      13.34638      34.817872 347.5911
## control             99.79867          35.60467      8.42000      8.993333 157.5840
##          Asparagine Betaine Carnitine Citrate Creatine Creatinine
## cachexic           75.39085          112.25298      64.62213     2720.853    174.91340 10722.140
## control             41.74900          55.97033      32.44367    1474.719    51.50433  5619.175
##          Dimethylamine Ethanolamine Formate Fucose Fumarate Glucose
## cachexic           453.5806          326.7721      187.56447     108.59915   10.92191 827.2189
## control             208.6833          197.1253      84.48333      57.44467   4.55200 140.9580
##          Glutamine Glycine Glycolate Guanidoacetate Hippurate Histidine
## cachexic           391.4104          1069.3779      219.2696       97.62426    2875.73 364.2323
## control             174.4273          585.1493      138.9837       68.73967    1364.24 180.4723
##          Hypoxanthine Isoleucine Lactate Leucine Lysine Methylamine
## cachexic           67.08702          9.660851      217.63191     31.26170    121.28234 21.21638
## control             51.71433          7.218000      65.74833      13.55667    89.22933 11.36000
##          Methylguanidine N.N.Dimethylglycine O.Acetyl carnitine Pantothenate
## cachexic           17.36468          34.48979       25.56447      39.94404
## control             12.12833          13.59667       10.59800      52.62267
##          Pyroglutamate Pyruvate Quinolinate Serine Succinate Sucrose
## cachexic           270.2923          26.86553       83.74723     245.8298    79.62894 150.02447
## control             119.2580          12.56633       39.32400     122.2630    29.83600 55.57967
##          Tartrate Taurine Threonine Trigonelline Trimethylamine.N.oxide
## cachexic           47.23468          655.7200      118.23319       359.6377    820.3406
## control             28.67600          320.5223      59.51867       130.6870    388.6690
##          Tryptophan Tyrosine Uracil Valine Xylose cis.Aconitate
## cachexic           81.82404          100.7423      37.51362     45.58255    129.28915 276.0255
## control             41.83300          52.0140      32.49333     20.13267    56.50933  91.7240
##          myo.Inositol trans.Aconitate pi.Methylhistidine tau.Methylhistidine
## cachexic           181.83766          48.81404       441.5532      105.66766
## control             62.64133          27.80933       258.6400      64.65033
```

```
apply(datos_cachexia, MARGIN = 2, FUN = tapply, INDEX=muscle, median)
```

```

##          X1.6.Anhydro.beta.D.glucose X1.Methylnicotinamide X2.Aminobutyrate
## cachexic                      68.720                  50.910                  15.18
## control                        34.985                  19.435                  7.58
##          X2.Hydroxyisobutyrate X2.Oxoglutarate X3.Aminoisobutyrate
## cachexic                      40.850                  67.360                  29.67
## control                        19.305                  31.685                  18.41
##          X3.Hydroxybutyrate X3.Hydroxyisovalerate X3.Indoxylsulfate
## cachexic                      19.490                  21.120                  188.67
## control                        6.655                   5.625                  105.11
##          X4.Hydroxyphenylacetate Acetate Acetone Adipate Alanine Asparagine
## cachexic                      86.49       70.81      7.920   15.030  304.90      62.80
## control                        50.70       17.32      6.925   6.295   78.65      29.52
##          Betaine Carnitine Citrate Creatine Creatinine Dimethylamine
## cachexic                      107.77      28.790     2230.54    76.710   9701.15      361.41
## control                        32.55       19.205     1043.62    19.515   3697.13      149.91
##          Ethanolamine Formate Fucose Fumarate Glucose Glutamine Glycine
## cachexic                      244.69      130.32     82.27      6.050   387.610     368.710     671.830
## control                        123.03      61.56      42.59      3.225   103.595     117.405     417.045
##          Glycolate Guanidoacetate Hippurate Histidine Hypoxanthine Isoleucine
## cachexic                      172.430      83.93     1919.850     284.290      43.82       8.17
## control                        70.605       48.00     602.595     88.375      33.67       4.31
##          Lactate Leucine Lysine Methylamine Methylguanidine N.N.Dimethylglycine
## cachexic                      106.700      23.10     96.54       19.30       8.85       28.79
## control                        40.885      9.12      35.34       5.34       6.82       8.94
##          O.Acetylcarnitine Pantothenate Pyroglutamate Pyruvate Quinolinate
## cachexic                      13.460       25.790      244.69      21.12      74.44
## control                        6.675       14.735      83.95       6.62      27.26
##          Serine Succinate Sucrose Tartrate Taurine Threonine Trigonelline
## cachexic                      206.44       45.150     58.560      15.96     336.97      98.49     186.790
## control                        98.28       11.805     19.395      10.75     192.21      39.45     72.275
##          Trimethylamine.N.oxide Tryptophan Tyrosine Uracil Valine Xylose
## cachexic                      482.99       79.04     85.630     29.370     38.47     70.11
## control                        257.70       22.69     27.005     21.355     13.47     34.88
##          cis.Aconitate myo.Inositol trans.Aconitate pi.Methylhistidine
## cachexic                      232.760      117.920      44.260      232.760
## control                        56.305      30.725      14.315      73.335
##          tau.Methylhistidine
## cachexic                      79.840
## control                        37.185

```

```
apply(datos_cachexia, MARGIN = 2, FUN = tapply, INDEX=muscle, sd)
```

```

##          X1.6.Anhydro.beta.D.glucose X1.Methylnicotinamide X2.Aminobutyrate
## cachexic                  142.43720           85.25987      33.861804
## control                   99.56823           186.96823      7.307203
##          X2.Hydroxyisobutyrate X2.Oxoglutarate X3.Aminoisobutyrate
## cachexic                  23.43735           411.9266       238.16424
## control                   21.99241           179.7103       57.84174
##          X3.Hydroxybutyrate X3.Hydroxyisovalerate X3.Indoxylsulfate
## cachexic                  30.734216          27.43863       211.6573
## control                   7.985001           16.99889       146.9752
##          X4.Hydroxyphenylacetate Acetate Acetone Adipate Alanine
## cachexic                  94.05853 90.60022 29.656970 62.18722 281.7285
## control                   154.80432 42.93314 5.652959 10.42638 156.1359
##          Asparagine Betaine Carnitine Citrate Creatine Creatinine
## cachexic      57.79106 83.19913 87.11959 2459.721 335.36486 6905.406
## control       40.19878 70.37315 40.45228 1313.056 87.23761 4228.759
##          Dimethylamine Ethanolamine Formate Fucose Fumarate Glucose
## cachexic      347.0764          264.5155 224.50099 91.12714 17.432383 1727.55510
## control       139.1947          211.0371 71.86832 47.17225 6.391354 99.19766
##          Glutamine Glycine Glycolate Guanidoacetate Hippurate Histidine
## cachexic      311.5721 1091.7293 186.3896          89.81720 3298.445 350.9996
## control       195.3615 570.1131 161.6588          69.08304 1700.305 199.0066
##          Hypoxanthine Isoleucine Lactate Leucine Lysine Methylamine
## cachexic      60.18018 7.572630 529.95631 24.231769 97.51977 13.78307
## control       52.72308 5.596286 55.81337 9.157233 162.20024 12.03258
##          Methylguanidine N.N.Dimethylglycine O.Acetylcarnitine Pantothenate
## cachexic      23.35677          26.59008          42.32996 41.09855
## control       12.23669          13.44479          10.28234 130.19429
##          Pyroglutamate Pyruvate Quinolinate Serine Succinate Sucrose
## cachexic      212.1925 29.62169 53.40643 214.4016 98.57757 315.0315
## control       98.6342 15.70713 33.76190 88.8920 44.99090 113.2921
##          Tartrate Taurine Threonine Trigonelline Trimethylamine.N.oxide
## cachexic      125.12896 786.4167 95.29957          476.9450          1099.2591
## control       54.87093 371.5426 57.42642          146.2309          369.4847
##          Tryptophan Tyrosine Uracil Valine Xylose cis.Aconitate
## cachexic      59.07686 94.08034 35.81805 32.48843 311.8836          331.28732
## control       42.08062 51.21703 34.05826 15.10151 80.2464          84.89998
##          myo.Inositol trans.Aconitate pi.Methylhistidine tau.Methylhistidine
## cachexic      197.76882          39.95090          612.8804          80.29638
## control       70.45362          35.95093          348.0329          65.91869

```

Después de evaluar de manera superficial los datos de cada metabolito debemos comprobar que los datos sigan una distribución normal (aunque se podría asumir puesto que la $n \geq 30$).

Para ello hacemos una función con el Shapiro test que nos devuelva la normalidad o no de cada uno de los grupos para cada uno de los metabolitos.

```
grupo_cachexic <- human_cachexia[human_cachexia$Muscle.loss == "cachexic", ]  
grupo_control <- human_cachexia[human_cachexia$Muscle.loss == "control", ]  
  
p_valores_normalidad <- sapply(names(datos_cachexia), function(metabolito) {  
  
    shapiro_cachexic <- shapiro.test(grupo_cachexic[[metabolito]])$p.value  
    shapiro_control <- shapiro.test(grupo_control[[metabolito]])$p.value  
    c(cachexic = shapiro_cachexic, control = shapiro_control)  
})  
  
p_valores_normalidad
```

```

##          X1.6.Anhydro.beta.D.glucose X1.Methylnicotinamide X2.Aminobutyrate
## cachexic           1.999866e-07      1.584013e-09     6.170993e-11
## control            1.613558e-08      1.631778e-10     9.541639e-04
##          X2.Hydroxyisobutyrate X2.Oxoglutarate X3.Aminoisobutyrate
## cachexic           0.039032782     2.303795e-12      1.465323e-12
## control            0.001626644     8.919452e-10      3.120749e-08
##          X3.Hydroxybutyrate X3.Hydroxyisovalerate X3.Indoxylsulfate
## cachexic           1.144397e-07     5.567290e-08      3.522801e-05
## control            1.960501e-04     1.123566e-07      4.252464e-06
##          X4.Hydroxyphenylacetate    Acetate     Acetone     Adipate
## cachexic           3.095323e-05     3.081564e-08     8.188739e-14     1.285528e-11
## control             6.849287e-09     2.386493e-06     7.606672e-05     2.078146e-08
##          Alanine   Asparagine   Betaine   Carnitine   Citrate
## cachexic           5.169099e-05     1.701573e-04     2.707096e-04     2.456377e-09     8.321229e-07
## control             6.674537e-05     4.929412e-06     1.011844e-06     2.278765e-07     2.235414e-03
##          Creatine   Creatinine  Dimethylamine Ethanolamine   Formate
## cachexic           5.864870e-12     0.012599185    0.0001238763    3.949457e-05     1.233095e-10
## control             3.317946e-08     0.002615974    0.0071950351    1.183419e-05     5.613942e-04
##          Fucose     Fumarate     Glucose   Glutamine   Glycine
## cachexic           0.0000361081    2.261986e-11    2.862457e-12    3.908030e-05     6.618303e-07
## control             0.0013534871    2.881270e-09    1.288603e-03    3.445112e-06     2.551701e-04
##          Glycolate  Guanidoacetate Hippurate   Histidine Hypoxanthine
## cachexic           1.699577e-04     2.886359e-08     2.314484e-08     9.460144e-07     1.121557e-05
## control             5.164470e-06     1.8666486e-05    6.496493e-07     2.765833e-05     9.258419e-05
##          Isoleucine  Lactate     Leucine   Lysine     Methylamine
## cachexic           1.289632e-07     1.580594e-13     7.102276e-06     3.859168e-05     4.354943e-03
## control             1.911635e-04     5.792578e-04     9.180866e-03     3.268827e-09     2.929615e-06
##          Methylguanidine N.N.Dimethylglycine O.Acetylcarnitine Pantothenate
## cachexic           5.426061e-10     0.0002357476    1.698556e-11     2.879223e-07
## control             3.359130e-05     0.0000823241    2.738320e-04     3.790773e-10
##          Pyroglutamate Pyruvate   Quinolinate   Serine     Succinate
## cachexic           9.497402e-05     7.209991e-09     1.534567e-03     2.913715e-07     4.172836e-09
## control             2.609957e-04     5.114396e-07     5.658773e-05     6.241602e-03     1.812575e-07
##          Sucrose     Tartrate     Taurine   Threonine Trigonelline
## cachexic           1.523150e-12     2.278197e-13    6.743389e-08     9.369478e-05     1.621765e-08
## control             6.125045e-10     3.952077e-09    1.035237e-05     4.538956e-05     5.288891e-06
##          Trimethylamine.N.oxide Tryptophan Tyrosine   Uracil
## cachexic           2.559041e-10     1.648129e-04    1.272778e-07     3.187664e-07
## control             4.925928e-05     2.994641e-06    6.269224e-05     7.906170e-06
##          Valine      Xylose   cis.Aconitate myo.Inositol trans.Aconitate
## cachexic           0.0003602307    7.416826e-14    3.54981e-09     1.009535e-07     1.745467e-05
## control             0.0007869125    5.004286e-09    1.96704e-04     2.153350e-06     5.209227e-08
##          pi.Methylhistidine tau.Methylhistidine
## cachexic           1.883894e-09     5.052134e-04
## control             1.051396e-06     5.480573e-05

```

Con los resultados <0.05 podemos asumir la normalidad de todos los parámetros.

Después deberemos comprobar si los datos de los metabolitos mantienen su homocedasticidad (homogeneidad de varianzas), factor muy importante cuando se trata de muestras con datos despareados (diferente n). Para ello realizaremos el test de Barlet.

```
p_valores_varianzas <- sapply(names(datos_cachexia), function(metabolito) {  
  bartlett.test(list(grupo_cachexic[[metabolito]], grupo_control[[metabolito]]))$p.value  
})  
  
p_valores_varianzas
```

## X1.6.Anhydro.beta.D.glucose	X1.Methylnicotinamide
## 4.078556e-02	2.667623e-06
## X2.Aminobutyrate	X2.Hydroxyisobutyrate
## 2.361929e-13	7.077151e-01
## X2.Oxoglutarate	X3.Aminoisobutyrate
## 9.226100e-06	5.555439e-12
## X3.Hydroxybutyrate	X3.Hydroxyisovalerate
## 3.273701e-11	7.067918e-03
## X3.Indoxylsulfate	X4.Hydroxyphenylacetate
## 3.734419e-02	2.675249e-03
## Acetate	Acetone
## 5.270733e-05	8.108063e-15
## Adipate	Alanine
## 2.368442e-16	1.076292e-03
## Asparagine	Betaine
## 3.820288e-02	3.285443e-01
## Carnitine	Citrate
## 3.463111e-05	5.426541e-04
## Creatine	Creatinine
## 3.380446e-11	5.877088e-03
## Dimethylamine	Ethanolamine
## 1.429341e-06	1.901399e-01
## Formate	Fucose
## 6.714089e-09	3.037699e-04
## Fumarate	Glucose
## 1.793217e-07	1.520834e-29
## Glutamine	Glycine
## 8.533954e-03	3.589739e-04
## Glycolate	Guanidoacetate
## 4.049773e-01	1.294174e-01
## Hippurate	Histidine
## 2.803221e-04	1.611793e-03
## Hypoxanthine	Isoleucine
## 4.385478e-01	8.196354e-02
## Lactate	Leucine
## 5.037986e-22	3.646063e-07
## Lysine	Methylamine
## 2.169188e-03	4.265723e-01
## Methylguanidine	N.N.Dimethylglycine
## 3.816691e-04	1.929635e-04
## O.Acetylcarnitine	Pantothenate
## 5.582396e-12	1.800147e-11
## Pyroglutamate	Pyruvate
## 3.541953e-05	4.788725e-04
## Quinolinate	Serine
## 9.685024e-03	3.015905e-06
## Succinate	Sucrose
## 2.415734e-05	1.134576e-07
## Tartrate	Taurine
## 1.030899e-05	4.955005e-05
## Threonine	Trigonelline
## 4.526319e-03	2.287092e-09
## Trimethylamine.N.oxide	Tryptophan
## 2.226998e-08	5.205522e-02
## Tyrosine	Uracil

```

##          7.789801e-04      7.663420e-01
##          Valine           Xylose
##          3.540937e-05      2.539016e-11
##          cis.Aconitate    myo.Inositol
##          2.287431e-11      9.060095e-08
##          trans.Aconitate   pi.Methylhistidine
##          5.357232e-01      1.656837e-03
##          tau.Methylhistidine
##          2.510004e-01

```

Comprobamos si alguno no mantiene la homogeneidad de las varianzas:

```

resultados_no_homocedasticidad <- p_valores_varianzas[p_valores_varianzas > 0.05]
resultados_no_homocedasticidad

```

```

## X2.Hydroxyisobutyrate      Betaine      Ethanolamine
## 0.70771510                 0.32854430  0.19013987
## Glycolate                  Guanidoacetate Hypoxanthine
## 0.40497734                 0.12941742  0.43854776
## Isoleucine                 Methylamine   Tryptophan
## 0.08196354                 0.42657234  0.05205522
## Uracil                     trans.Aconitate tau.Methylhistidine
## 0.76634204                 0.53572316  0.25100042

```

Como hemos encontrado valores que no conservan la homocedasticidad deberemos realizar una t de Student con la corrección de Welch para buscar las diferencias en cada uno de los grupos e intentaremos observar que metabolitos tienen diferencias significativas.

Como se suele hacer:

H0: No hay diferencias significativas entre las medias de los 2 grupos, por lo que no podemos asumir que las diferencias observadas no sean fruto del azar.

H1: hay diferencias significativas entre las medias de los 2 grupos, por lo que podemos asumir que las diferencias observadas se deben a los factores dentro de los grupos.

```

resultados_ttest <- sapply(names(datos_cachexia), function(metabolito) {
  t.test(grupo_cachexic[[metabolito]], grupo_control[[metabolito]], var.equal=FALSE)$p.value
})
resultados_ttest

```

## X1.6.Anhydro.beta.D.glucose	X1.Methylnicotinamide
## 3.531943e-02	9.435367e-01
## X2.Aminobutyrate	X2.Hydroxyisobutyrate
## 7.859048e-03	4.893295e-03
## X2.Oxoglutarate	X3.Aminoisobutyrate
## 1.585788e-01	1.021946e-01
## X3.Hydroxybutyrate	X3.Hydroxyisovalerate
## 1.334590e-04	3.458034e-03
## X3.Indoxylsulfate	X4.Hydroxyphenylacetate
## 4.849645e-03	5.272919e-01
## Acetate	Acetone
## 1.740727e-03	2.731685e-01
## Adipate	Alanine
## 7.515595e-03	2.960803e-04
## Asparagine	Betaine
## 3.566603e-03	2.174920e-03
## Carnitine	Citrate
## 3.192506e-02	5.098507e-03
## Creatine	Creatinine
## 1.984702e-02	1.369605e-04
## Dimethylamine	Ethanolamine
## 5.348990e-05	2.011237e-02
## Formate	Fucose
## 4.909304e-03	1.863466e-03
## Fumarate	Glucose
## 2.620524e-02	9.239445e-03
## Glutamine	Glycine
## 3.391600e-04	1.304117e-02
## Glycolate	Guanidoacetate
## 4.940718e-02	1.165822e-01
## Hippurate	Histidine
## 1.015460e-02	4.541154e-03
## Hypoxanthine	Isoleucine
## 2.421171e-01	1.087749e-01
## Lactate	Leucine
## 5.733204e-02	2.662822e-05
## Lysine	Methylamine
## 3.347563e-01	1.496975e-03
## Methylguanidine	N.N.Dimethylglycine
## 2.027750e-01	2.114015e-05
## O.Acetylcarnitine	Pantothenate
## 2.418993e-02	6.085017e-01
## Pyroglutamate	Pyruvate
## 7.289814e-05	7.357362e-03
## Quinolinate	Serine
## 2.708732e-05	8.173888e-04
## Succinate	Sucrose
## 3.676506e-03	6.558505e-02
## Tartrate	Taurine
## 3.758724e-01	1.418748e-02
## Threonine	Trigonelline
## 1.183249e-03	3.219917e-03
## Trimethylamine.N.oxide	Tryptophan
## 1.586788e-02	8.885986e-04
## Tyrosine	Uracil

```
##          4.456642e-03      5.386741e-01
##          Valine           Xylose
##          1.574212e-05     1.335317e-01
##          cis.Aconitate   myo.Inositol
##          6.190184e-04     3.611776e-04
##          trans.Aconitate pi.Methylhistidine
##          1.953314e-02    9.958934e-02
##          tau.Methylhistidine
##          1.710741e-02
```

Aplicamos el test de Welch con var.equal=FALSE y puesto que el resultado de la comparativa de la t de student es bastante extensa con \$p.value hacemos que solo nos devuelva el valor del p.valor para simplificar la obtención de los resultados.

Del mismo modo podemos obtener un listado con los metabolitos que presenten diferencias significativas y los que no.

```
resultados_significativos <- resultados_ttest[resultados_ttest < 0.05]
resultados_significativos
```

```

## X1.6.Anhydro.beta.D.glucose          X2.Aminobutyrate
##                               3.531943e-02    7.859048e-03
## X2.Hydroxyisobutyrate               X3.Hydroxybutyrate
##                               4.893295e-03    1.334590e-04
## X3.Hydroxyisovalerate              X3.Indoxylsulfate
##                               3.458034e-03    4.849645e-03
## Acetate                           Adipate
##                               1.740727e-03    7.515595e-03
## Alanine                           Asparagine
##                               2.960803e-04    3.566603e-03
## Betaine                           Carnitine
##                               2.174920e-03    3.192506e-02
## Citrate                           Creatine
##                               5.098507e-03    1.984702e-02
## Creatinine                         Dimethylamine
##                               1.369605e-04    5.348990e-05
## Ethanolamine                       Formate
##                               2.011237e-02    4.909304e-03
## Fucose                            Fumarate
##                               1.863466e-03    2.620524e-02
## Glucose                           Glutamine
##                               9.239445e-03    3.391600e-04
## Glycine                           Glycolate
##                               1.304117e-02    4.940718e-02
## Hippurate                          Histidine
##                               1.015460e-02    4.541154e-03
## Leucine                            Methylamine
##                               2.662822e-05    1.496975e-03
## N.N.Dimethylglycine              O.Acetylcarnitine
##                               2.114015e-05    2.418993e-02
## Pyroglutamate                      Pyruvate
##                               7.289814e-05    7.357362e-03
## Quinolinate                        Serine
##                               2.708732e-05    8.173888e-04
## Succinate                          Taurine
##                               3.676506e-03    1.418748e-02
## Threonine                          Trigonelline
##                               1.183249e-03    3.219917e-03
## Trimethylamine.N.oxide            Tryptophan
##                               1.586788e-02    8.885986e-04
## Tyrosine                           Valine
##                               4.456642e-03    1.574212e-05
## cis.Aconitate                      myo.Inositol
##                               6.190184e-04    3.611776e-04
## trans.Aconitate                   tau.Methylhistidine
##                               1.953314e-02    1.710741e-02

```

```

resultados_no_significativos <- resultados_ttest[resultados_ttest > 0.05]
resultados_no_significativos

```

## X1.Methylnicotinamide	X2.Oxoglutarate	X3.Aminoisobutyrate
## 0.94353674	0.15857875	0.10219457
## X4.Hydroxyphenylacetate	Acetone	Guanidoacetate
## 0.52729195	0.27316846	0.11658217
## Hypoxanthine	Isoleucine	Lactate
## 0.24211709	0.10877489	0.05733204
## Lysine	Methylguanidine	Pantothenate
## 0.33475631	0.20277495	0.60850167
## Sucrose	Tartrate	Uracil
## 0.06558505	0.37587240	0.53867410
## Xylose	pi.Methylhistidine	
## 0.13353169	0.09958934	

Y generar un listado solo con el nombre de los metabolitos significativos.

```
metabolitos_significativos <- names(resultados_ttest)[resultados_ttest < 0.05]
metabolitos_significativos
```

## [1] "X1.6.Anhydro.beta.D.glucose"	"X2.Aminobutyrate"
## [3] "X2.Hydroxyisobutyrate"	"X3.Hydroxybutyrate"
## [5] "X3.Hydroxyisovalerate"	"X3.Indoxylsulfate"
## [7] "Acetate"	"Adipate"
## [9] "Alanine"	"Asparagine"
## [11] "Betaine"	"Carnitine"
## [13] "Citrate"	"Creatine"
## [15] "Creatinine"	"Dimethylamine"
## [17] "Ethanolamine"	"Formate"
## [19] "Fucose"	"Fumarate"
## [21] "Glucose"	"Glutamine"
## [23] "Glycine"	"Glycolate"
## [25] "Hippurate"	"Histidine"
## [27] "Leucine"	"Methylamine"
## [29] "N.N.Dimethylglycine"	"O.Acetylcarnitine"
## [31] "Pyroglutamate"	"Pyruvate"
## [33] "Quinolinate"	"Serine"
## [35] "Succinate"	"Taurine"
## [37] "Threonine"	"Trigonelline"
## [39] "Trimethylamine.N.oxide"	"Tryptophan"
## [41] "Tyrosine"	"Valine"
## [43] "cis.Aconitate"	"myo.Inositol"
## [45] "trans.Aconitate"	"tau.Methylhistidine"

5- Resultados

Con este listado podríamos contestar a la pregunta:

¿hay diferencias significativas entre control y cachexia para la cantidad de alguno de los metabolitos presentes en orina?

Si. Hay diferencias significativas.

Si asumimos que la función de este estudio tiene un perfil orientado al diagnostico del síndrome de caquexia se podría utilizar una combinación de ellos para definir un panel de diagnostico que permitiese su utilización en un contexto hospitalario.

La combinación final de metabolitos deberá ser evaluada en función de otras consideraciones como: su detección, su presencia en otras enfermedades o la factibilidad de ser implementado en un laboratorio hospitalario, entre muchas otras.

6- URL repositorio Github

<https://github.com/SergioHHTT/Huertas-Torres-Sergio-PEC1> (<https://github.com/SergioHHTT/Huertas-Torres-Sergio-PEC1>)