PAC2

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Índice

El objetivo de esta PEC es ilustrar el proceso de análisis de datos de ultrasecuenciación mediante la realización de un estudio, tal como se llevaria a cabo en una situación real. La PEC se basa en los datos suministrados de un estudio del que se debe extraer una muestra aleatoria con el fin de garantizar que cada conjunto de datos sea distinto.

1. Entorno de trabajo y lectura de ficheros

Creamos el entorno de trabajo y leemos el fichero targets.csv

```
setwd("C:/Users/CarlesM/Desktop/pac2")
getwd()
```

```
## [1] "C:/Users/CarlesM/Desktop/pac2"
```

```
targets<- read.csv2(file.path("./data", "targets.csv"), head=T, sep=",")
head(targets,5)
```

```
##   Experiment SRA_Sample              Sample_Name Grupo_analisis body_site
## 1  SRX567480   SRS626942 GTEX-111CU-0226-SM-5GZXC              1   Thyroid
## 2  SRX615964   SRS644174 GTEX-111FC-1026-SM-5GZX1              1   Thyroid
## 3  SRX563960   SRS625636 GTEX-111VG-0526-SM-5N9BW              3   Thyroid
## 4  SRX564185   SRS625665 GTEX-111YS-0726-SM-5GZY8              1   Thyroid
## 5  SRX559141   SRS624025 GTEX-1122O-0226-SM-5N9DA              1   Thyroid
##          molecular_data_type    sex Group ShortName
## 1 Allele-Specific Expression   male   NIT 111CU_NIT
## 2             RNA Seq (NGS)   male   NIT 111FC_NIT
## 3             RNA Seq (NGS)   male   ELI 111VG_ELI
## 4 Allele-Specific Expression   male   NIT 111YS_NIT
## 5             RNA Seq (NGS) female   NIT 1122O_NIT
```

2. Elección de datos de cada grupo de targets

Escogemos los datos de cada grupo del archivo targets

```
datos_NIT<-targets[targets$Group=="NIT",]
head(datos_NIT,5)
```

```
##   Experiment SRA_Sample              Sample_Name Grupo_analisis body_site
## 1  SRX567480   SRS626942 GTEX-111CU-0226-SM-5GZXC              1   Thyroid
## 2  SRX615964   SRS644174 GTEX-111FC-1026-SM-5GZX1              1   Thyroid
## 4  SRX564185   SRS625665 GTEX-111YS-0726-SM-5GZY8              1   Thyroid
## 5  SRX559141   SRS624025 GTEX-1122O-0226-SM-5N9DA              1   Thyroid
## 6  SRX561718   SRS625313 GTEX-1128S-0126-SM-5H12S              1   Thyroid
##          molecular_data_type    sex Group ShortName
## 1 Allele-Specific Expression   male   NIT 111CU_NIT
## 2             RNA Seq (NGS)   male   NIT 111FC_NIT
## 4 Allele-Specific Expression   male   NIT 111YS_NIT
## 5             RNA Seq (NGS) female   NIT 1122O_NIT
## 6 Allele-Specific Expression female   NIT 1128S_NIT
```

```
datos_ELI<-targets[targets$Group=="ELI",]
head(datos_ELI,5)
```

```
##     Experiment SRA_Sample              Sample_Name Grupo_analisis body_site
## 3    SRX563960   SRS625636 GTEX-111VG-0526-SM-5N9BW              3   Thyroid
## 29   SRX628009   SRS648152 GTEX-11NV4-0626-SM-5N9BR              3   Thyroid
## 40   SRX619829   SRS644736 GTEX-11XUK-0226-SM-5EQLW              3   Thyroid
## 100  SRX582762   SRS631169 GTEX-13NZ9-1126-SM-5MR37              3   Thyroid
## 119  SRX601511   SRS638114 GTEX-13QJC-0826-SM-5RQKC              3   Thyroid
##            molecular_data_type    sex Group ShortName
## 3               RNA Seq (NGS)   male   ELI 111VG_ELI
## 29              RNA Seq (NGS)   male   ELI 11NV4_ELI
## 40              RNA Seq (NGS) female   ELI 11XUK_ELI
## 100             RNA Seq (NGS)   male   ELI 13NZ9_ELI
## 119 Allele-Specific Expression female   ELI 13QJC_ELI
```

```
datos_SFI<-targets[targets$Group=="SFI",]
head(datos_SFI,5)
```

```
##    Experiment SRA_Sample          Sample_Name Grupo_analisis body_site
## 9   SRX557750  SRS623875 GTEX-117YW-0126-SM-5EGGN          2   Thyroid
## 14  SRX578169  SRS629611 GTEX-11DXY-0426-SM-5H12R          2   Thyroid
## 21  SRX619524  SRS644703 GTEX-11EQ8-0826-SM-5N9FG          2   Thyroid
## 22  SRX558144  SRS623916 GTEX-11EQ9-0626-SM-5A5K1          2   Thyroid
## 23  SRX567902  SRS627040 GTEX-11GS4-0826-SM-5986J          2   Thyroid
##              molecular_data_type  sex Group ShortName
## 9               RNA Seq (NGS) male   SFI 117YW_SFI
## 14              RNA Seq (NGS) male   SFI 11DXY_SFI
## 21 Allele-Specific Expression male   SFI 11EQ8_SFI
## 22              RNA Seq (NGS) male   SFI 11EQ9_SFI
## 23              RNA Seq (NGS) male   SFI 11GS4_SFI
```

3. Extracción aleatoria de muestras de cada grupo

Extraemos las 10 muestras aleatoriamente de cada grupo

```
muestra.NIT = datos_NIT[sample(nrow(datos_NIT),10) , ]
muestra.SFI = datos_SFI[sample(nrow(datos_SFI),10) , ]
muestra.ELI = datos_ELI[sample(nrow(datos_ELI),10) , ]

muestra.NIT
```

```
##       Experiment SRA_Sample          Sample_Name Grupo_analisis body_site
## 155    SRX199228  SRS333389  GTEX-OHPK-2626-SM-2HMK9          1   Thyroid
## 32     SRX578920  SRS629748 GTEX-11P7K-0226-SM-5986Z          1   Thyroid
## 125    SRX607930  SRS639555 GTEX-13W46-0926-SM-5LU3T          1   Thyroid
## 231    SRX411996  SRS524824  GTEX-XLM4-0726-SM-4AT64          1   Thyroid
## 118    SRX601210  SRS638072 GTEX-13QBU-0626-SM-5J2OG          1   Thyroid
## 209    SRX221407  SRS389066  GTEX-T8EM-0226-SM-3DB7C          1   Thyroid
## 181    SRX203754  SRS374845  GTEX-QXCU-0326-SM-2TC63          1   Thyroid
## 65     SRX632358  SRS648643 GTEX-131XE-0126-SM-5LZVC          1   Thyroid
## 81     SRX623670  SRS645895 GTEX-13CF3-0926-SM-5LZZC          1   Thyroid
## 201    SRX633885  SRS648815  GTEX-S7SF-0226-SM-5SI7H          1   Thyroid
##              molecular_data_type    sex Group ShortName
## 155 Allele-Specific Expression female   NIT OHPK-_NIT
## 32              RNA Seq (NGS)   male   NIT 11P7K_NIT
## 125             RNA Seq (NGS)   male   NIT 13W46_NIT
## 231             RNA Seq (NGS)   male   NIT XLM4-_NIT
## 118             RNA Seq (NGS) female   NIT 13QBU_NIT
## 209             RNA Seq (NGS)   male   NIT T8EM-_NIT
## 181             RNA Seq (NGS)   male   NIT QXCU-_NIT
## 65              RNA Seq (NGS)   male   NIT 131XE_NIT
## 81              RNA Seq (NGS) female   NIT 13CF3_NIT
## 201 Allele-Specific Expression female   NIT S7SF-_NIT
```

```
muestra.SFI
```

```
##      Experiment SRA_Sample            Sample_Name Grupo_analisis body_site
## 102   SRX614680   SRS644012 GTEX-13O1R-0826-SM-5J2MB            2   Thyroid
## 37    SRX616143   SRS644194 GTEX-11TUW-0226-SM-5LU8X            2   Thyroid
## 232   SRX407488   SRS524425  GTEX-XMK1-0626-SM-4B65A            2   Thyroid
## 166   SRX198253   SRS333514  GTEX-P78B-0526-SM-2I5F7            2   Thyroid
## 58    SRX601535   SRS638117 GTEX-12ZZX-1226-SM-5EGHS            2   Thyroid
## 210   SRX222167   SRS389523  GTEX-TKQ1-0126-SM-33HB3            2   Thyroid
## 9     SRX557750   SRS623875 GTEX-117YW-0126-SM-5EGGN            2   Thyroid
## 171   SRX199518   SRS333107  GTEX-Q2AH-0726-SM-2I3EA            2   Thyroid
## 203   SRX223043   SRS389833  GTEX-SIU8-0626-SM-2XCDN            2   Thyroid
## 178   SRX198185   SRS333004  GTEX-QLQ7-0726-SM-2I5G2            2   Thyroid
##          molecular_data_type    sex Group ShortName
## 102            RNA Seq (NGS)    male   SFI 13O1R_SFI
## 37             RNA Seq (NGS)    male   SFI 11TUW_SFI
## 232            RNA Seq (NGS)    male   SFI XMK1-_SFI
## 166 Allele-Specific Expression female   SFI P78B-_SFI
## 58             RNA Seq (NGS) female   SFI 12ZZX_SFI
## 210            RNA Seq (NGS)    male   SFI TKQ1-_SFI
## 9              RNA Seq (NGS)    male   SFI 117YW_SFI
## 171 Allele-Specific Expression    male   SFI Q2AH-_SFI
## 203 Allele-Specific Expression    male   SFI SIU8-_SFI
## 178            RNA Seq (NGS)    male   SFI QLQ7-_SFI
```

```
muestra.ELI
```

```
##      Experiment SRA_Sample            Sample_Name Grupo_analisis body_site
## 40    SRX619829   SRS644736 GTEX-11XUK-0226-SM-5EQLW            3   Thyroid
## 211   SRX222429   SRS389623  GTEX-TMMY-0826-SM-33HB9            3   Thyroid
## 3     SRX563960   SRS625636 GTEX-111VG-0526-SM-5N9BW            3   Thyroid
## 146   SRX575932   SRS629299 GTEX-14ABY-0926-SM-5Q5DY            3   Thyroid
## 253   SRX583148   SRS631283  GTEX-YJ89-0726-SM-5P9F7            3   Thyroid
## 290   SRX568364   SRS627095  GTEX-ZYY3-1926-SM-5GZXS            3   Thyroid
## 147   SRX607358   SRS639491 GTEX-14AS3-0226-SM-5Q5B6            3   Thyroid
## 100   SRX582762   SRS631169 GTEX-13NZ9-1126-SM-5MR37            3   Thyroid
## 186   SRX204036   SRS374975  GTEX-R55G-0726-SM-2TC6J            3   Thyroid
## 167   SRX199272   SRS333099  GTEX-PLZ4-1226-SM-2I5FE            3   Thyroid
##          molecular_data_type    sex Group ShortName
## 40             RNA Seq (NGS) female   ELI 11XUK_ELI
## 211 Allele-Specific Expression female   ELI TMMY-_ELI
## 3              RNA Seq (NGS)    male   ELI 111VG_ELI
## 146 Allele-Specific Expression    male   ELI 14ABY_ELI
## 253            RNA Seq (NGS)    male   ELI YJ89-_ELI
## 290 Allele-Specific Expression female   ELI ZYY3-_ELI
## 147            RNA Seq (NGS) female   ELI 14AS3_ELI
## 100            RNA Seq (NGS)    male   ELI 13NZ9_ELI
## 186            RNA Seq (NGS) female   ELI R55G-_ELI
## 167            RNA Seq (NGS) female   ELI PLZ4-_ELI
```

4. Coincidencias con el archivo counts.csv

Seleccionamos las columnas de counts que coincidan con la columna Sample-Name de los 30 targets y leemos el archivo resultante scounts. La elección de las columnas se ha hecho usando Excel

```
scounts<- read.csv2(file.path("./data", "selectcounts.csv"), head=T, sep=";")
str(scounts)
```

```
## 'data.frame':    56202 obs. of  31 variables:
##  $ X                  : Factor w/ 56202 levels "ENSG00000000003.10",..: 26352
28704 39144 36095 53325 37828 16363 36388 33329 36152 ...
##  $ GTEX.111VG.0526.SM.5N9BW: int  1 474 1 0 1 1 0 3 7 427 ...
##  $ GTEX.11EM3.0126.SM.5985K: int  2 669 2 1 1 1 0 3 20 791 ...
##  $ GTEX.11EMC.0226.SM.5EGLP: int  5 786 0 0 0 1 0 10 8 553 ...
##  $ GTEX.11NSD.0126.SM.5987F: int  0 408 1 0 0 0 2 11 19 800 ...
##  $ GTEX.11NV4.0626.SM.5N9BR: int  3 1301 1 0 0 1 0 5 7 1132 ...
##  $ GTEX.11O72.2326.SM.5BC7H: int  0 633 2 1 0 1 1 14 11 1075 ...
##  $ GTEX.12WSG.0226.SM.5EGIF: int  3 369 1 3 1 2 2 3 10 235 ...
##  $ GTEX.139UW.0126.SM.5KM1B: int  2 430 0 0 0 0 0 9 9 679 ...
##  $ GTEX.13NZ9.1126.SM.5MR37: int  0 1002 1 0 0 1 0 15 19 602 ...
##  $ GTEX.13O1R.0826.SM.5J2MB: int  3 460 0 1 2 0 1 7 12 279 ...
##  $ GTEX.13OVG.0226.SM.5LU93: int  4 719 2 1 2 2 1 6 14 1064 ...
##  $ GTEX.13QJC.0826.SM.5RQKC: int  0 825 1 0 0 1 1 10 21 853 ...
##  $ GTEX.13U4I.0526.SM.5LU59: int  2 636 0 0 0 0 0 8 13 606 ...
##  $ GTEX.14ABY.0926.SM.5Q5DY: int  1 775 2 0 0 0 1 10 2 580 ...
##  $ GTEX.14AS3.0226.SM.5Q5B6: int  0 834 1 1 0 0 0 6 6 445 ...
##  $ GTEX.14BMU.0226.SM.5S2QA: int  2 423 0 0 2 1 0 18 6 325 ...
##  $ GTEX.PWN1.2626.SM.2I3FH : int  5 297 0 0 1 2 0 0 453 229 ...
##  $ GTEX.S7SE.0726.SM.2XCD7 : int  4 422 0 1 1 2 1 4 12 247 ...
##  $ GTEX.T5JW.1226.SM.3GACY : int  1 541 2 0 0 0 1 1 9 1468 ...
##  $ GTEX.WYVS.0326.SM.3NM9V : int  6 820 0 1 0 4 5 12 18 973 ...
##  $ GTEX.XBED.0126.SM.47JY7 : int  3 766 3 4 0 4 1 10 11 374 ...
##  $ GTEX.XMK1.0626.SM.4B65A : int  9 568 1 1 1 0 1 5 14 738 ...
##  $ GTEX.Y5V6.0526.SM.4VBRV : int  3 482 3 2 2 2 2 2 27 681 ...
##  $ GTEX.YEC4.0626.SM.5CVLU : int  1 365 1 1 0 1 1 1 20 359 ...
##  $ GTEX.YFC4.2626.SM.5P9FQ : int  1 1472 1 0 0 1 2 38 24 2020 ...
##  $ GTEX.YJ89.0726.SM.5P9F7 : int  4 1325 1 0 2 1 2 4 8 853 ...
##  $ GTEX.Z9EW.0226.SM.5CVM7 : int  3 450 2 2 0 1 0 2 10 352 ...
##  $ GTEX.ZLV1.0126.SM.4WWBZ : int  2 689 2 4 0 2 0 18 9 809 ...
##  $ GTEX.ZYVF.1126.SM.5E458 : int  2 838 1 4 1 1 0 0 21 1212 ...
##  $ GTEX.ZYY3.1926.SM.5GZXS : int  6 1003 1 2 0 1 4 8 12 960 ...
```

5. Packages

Ahora tendremos que cargar los paquetes que necesitaremos

```
library(edgeR)
```

```
## Loading required package: limma
```

```
library (limma)
library(Glimma)
library(gplots)
```

```
## Warning: package 'gplots' was built under R version 3.6.3
```

```
##
## Attaching package: 'gplots'
```

```
## The following object is masked from 'package:stats':
##
##     lowess
```

```
library(org.Mm.eg.db)
```

```
## Loading required package: AnnotationDbi
```

```
## Loading required package: stats4
```

```
## Loading required package: BiocGenerics
```

```
## Loading required package: parallel
```

```
##
## Attaching package: 'BiocGenerics'
```

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

```
## The following object is masked from 'package:limma':
##
##     plotMA
```

```
## The following objects are masked from 'package:stats':
##
##     IQR, mad, sd, var, xtabs
```

```
## The following objects are masked from 'package:base':
##
##     anyDuplicated, append, as.data.frame, basename, cbind, colnames,
##     dirname, do.call, duplicated, eval, evalq, Filter, Find, get, grep,
##     grepl, intersect, is.unsorted, lapply, Map, mapply, match, mget,
##     order, paste, pmax, pmax.int, pmin, pmin.int, Position, rank,
##     rbind, Reduce, rownames, sapply, setdiff, sort, table, tapply,
##     union, unique, unsplit, which, which.max, which.min
```

```
## Loading required package: Biobase
```

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

```
## Loading required package: IRanges
```

```
## Loading required package: S4Vectors
```

```
## Warning: package 'S4Vectors' was built under R version 3.6.3
```

```
##
## Attaching package: 'S4Vectors'
```

```
## The following object is masked from 'package:gplots':
##
##     space
```

```
## The following object is masked from 'package:base':
##
##     expand.grid
```

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

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

```
##
```

```
library(RColorBrewer)
library(DESeq2)
```

```
## Loading required package: GenomicRanges
```

```
## Loading required package: GenomeInfoDb
```

```
## Warning: package 'GenomeInfoDb' was built under R version 3.6.3
```

```
## Loading required package: SummarizedExperiment
```

```
## Loading required package: DelayedArray
```

```
## Warning: package 'DelayedArray' was built under R version 3.6.3
```

```
## Loading required package: matrixStats
```

```
## Warning: package 'matrixStats' was built under R version 3.6.3
```

```
##
## Attaching package: 'matrixStats'
```

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

```
## Loading required package: BiocParallel
```

```
##
## Attaching package: 'DelayedArray'
```

```
## The following objects are masked from 'package:matrixStats':
##
##     colMaxs, colMins, colRanges, rowMaxs, rowMins, rowRanges
```

```
## The following objects are masked from 'package:base':
##
##     aperm, apply, rowsum
```

6. Lectura de datos y filtrado y eliminacion de genes con contajes bajos

Los genes con recuentos muy bajos en todas las bibliotecas proporcionan poca evidencia en la expresión diferencial e interfieren con algunas de las aproximaciones estadísticas que se utilizan más adelante dentro del pipeleine del análisis.

Asimismo añaden "ruido"" en el ajuste por múltiple testing mediante FDR, reduciendo "potencia estadística" en la detección de genes expresados diferencialmente (como ya hemos discutido en debates anteriores).

Estos genes deben filtrarse antes de un análisis posterior.

Hay diferentes maneras de filtrar genes poco expresados. En este caso optamos por retener los genes si se expresan en un conteo por millón (CPM) por encima de 0.5 en al menos dos muestras.

Utilizaremos la función cpm del package edgeR para generar los valores de CPM y luego filtrar. Hay que tener presente que al convertir a CPM estamos normalizando segun el "Sequencing depth" de cada muestra.

Nota: Secuencing depth es comúnmente un término usado para la secuenciación del genoma o del exoma y significa el número de lecturas que cubren cada posición.

```
rownames(scounts)<-scounts[,1]
scounts<-scounts[,-(1)]
library(edgeR)
dgeList_counts<-DGEList(scounts)
counts_cpm<-cpm(dgeList_counts,log=TRUE)
head(counts_cpm)
```

```
##                     GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000223972.4                 -4.162194                 -3.890311
## ENSG00000227232.4                  3.188591                  3.367278
## ENSG00000243485.2                 -4.162194                 -3.890311
## ENSG00000237613.2                 -4.768198                 -4.263485
## ENSG00000268020.2                 -4.162194                 -4.263485
## ENSG00000240361.1                 -4.162194                 -4.263485
##                     GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000223972.4                 -3.149473                 -4.768198
## ENSG00000227232.4                  3.583036                  2.973675
## ENSG00000243485.2                 -4.768198                 -4.162034
## ENSG00000237613.2                 -4.768198                 -4.768198
## ENSG00000268020.2                 -4.768198                 -4.768198
## ENSG00000240361.1                 -4.268205                 -4.768198
##                     GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000223972.4                 -3.404865                 -4.768198
## ENSG00000227232.4                  4.647684                  3.214006
## ENSG00000243485.2                 -4.160086                 -3.923594
## ENSG00000237613.2                 -4.768198                 -4.284952
## ENSG00000268020.2                 -4.768198                 -4.768198
## ENSG00000240361.1                 -4.160086                 -4.284952
##                     GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000223972.4                 -3.128109                 -3.676954
## ENSG00000227232.4                  3.261766                  3.162977
## ENSG00000243485.2                 -3.997905                 -4.768198
## ENSG00000237613.2                 -3.128109                 -4.768198
## ENSG00000268020.2                 -3.997905                 -4.768198
## ENSG00000240361.1                 -3.498427                 -4.768198
##                     GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000223972.4                 -4.768198                 -3.538156
## ENSG00000227232.4                  4.030623                  2.927702
## ENSG00000243485.2                 -4.238656                 -4.768198
## ENSG00000237613.2                 -4.768198                 -4.233559
## ENSG00000268020.2                 -4.768198                 -3.844358
## ENSG00000240361.1                 -4.238656                 -4.768198
##                     GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000223972.4                 -3.039814                 -4.768198
## ENSG00000227232.4                  3.935232                  4.081484
## ENSG00000243485.2                 -3.659318                 -4.128516
## ENSG00000237613.2                 -4.109743                 -4.768198
## ENSG00000268020.2                 -3.659318                 -4.768198
## ENSG00000240361.1                 -3.659318                 -4.128516
##                     GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000223972.4                 -3.630268                 -4.261134
## ENSG00000227232.4                  3.812274                  3.586691
## ENSG00000243485.2                 -4.768198                 -3.886682
## ENSG00000237613.2                 -4.768198                 -4.768198
## ENSG00000268020.2                 -4.768198                 -4.768198
## ENSG00000240361.1                 -4.768198                 -4.768198
##                     GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000223972.4                 -4.768198                 -3.614303
## ENSG00000227232.4                  4.313859                  3.254818
## ENSG00000243485.2                 -4.046922                 -4.768198
## ENSG00000237613.2                 -4.046922                 -4.768198
## ENSG00000268020.2                 -4.768198                 -3.614303
## ENSG00000240361.1                 -4.768198                 -4.078837
##                     GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
```

```
## ENSG00000223972.4                 -2.620847                 -2.922711
## ENSG00000227232.4                  2.909522                  3.333199
## ENSG00000243485.2                 -4.768198                 -4.768198
## ENSG00000237613.2                 -4.768198                 -4.047100
## ENSG00000268020.2                 -4.014570                 -4.047100
## ENSG00000240361.1                 -3.522061                 -3.568528
##                        GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
## ENSG00000223972.4                 -4.101799                 -3.121589
## ENSG00000227232.4                  3.547515                  3.422774
## ENSG00000243485.2                 -3.647705                 -4.768198
## ENSG00000237613.2                 -4.768198                 -4.329733
## ENSG00000268020.2                 -4.768198                 -4.768198
## ENSG00000240361.1                 -4.768198                 -3.492806
##                        GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000223972.4                 -3.316634                 -2.275165
## ENSG00000227232.4                  4.026265                  3.427442
## ENSG00000243485.2                 -3.316634                 -4.169451
## ENSG00000237613.2                 -3.039889                 -4.169451
## ENSG00000268020.2                 -4.768198                 -4.169451
## ENSG00000240361.1                 -3.039889                 -4.768198
##                        GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000223972.4                 -3.637457                 -4.147308
## ENSG00000227232.4                  2.817869                  2.856089
## ENSG00000243485.2                 -3.637457                 -4.147308
## ENSG00000237613.2                 -3.925709                 -4.147308
## ENSG00000268020.2                 -3.925709                 -4.768198
## ENSG00000240361.1                 -3.925709                 -4.147308
##                        GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000223972.4                 -4.350836                 -3.461805
## ENSG00000227232.4                  4.182595                  4.165509
## ENSG00000243485.2                 -4.350836                 -4.315807
## ENSG00000237613.2                 -4.768198                 -4.768198
## ENSG00000268020.2                 -4.768198                 -3.971922
## ENSG00000240361.1                 -4.350836                 -4.315807
##                        GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000223972.4                 -3.310837                 -3.780305
## ENSG00000227232.4                  3.270219                  3.640098
## ENSG00000243485.2                 -3.654472                 -3.780305
## ENSG00000237613.2                 -3.654472                 -3.199401
## ENSG00000268020.2                 -4.768198                 -4.768198
## ENSG00000240361.1                 -4.106431                 -3.780305
##                        GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000223972.4                 -3.657650                 -2.649670
## ENSG00000227232.4                  4.158799                  4.360455
## ENSG00000243485.2                 -4.108604                 -4.129351
## ENSG00000237613.2                 -3.037644                 -3.688108
## ENSG00000268020.2                 -4.108604                 -4.768198
## ENSG00000240361.1                 -4.108604                 -4.129351
```

```
# Which values in myCPM are greater than 0.5?
thresh <- counts_cpm > 0.5
# This produces a logical matrix with TRUEs and FALSEs
head(thresh)
```

```
##                      GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000223972.4                      FALSE                    FALSE
## ENSG00000227232.4                       TRUE                     TRUE
## ENSG00000243485.2                      FALSE                    FALSE
## ENSG00000237613.2                      FALSE                    FALSE
## ENSG00000268020.2                      FALSE                    FALSE
## ENSG00000240361.1                      FALSE                    FALSE
##                      GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
```

```
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
##                       GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000223972.4                       FALSE                    FALSE
## ENSG00000227232.4                        TRUE                     TRUE
## ENSG00000243485.2                       FALSE                    FALSE
## ENSG00000237613.2                       FALSE                    FALSE
## ENSG00000268020.2                       FALSE                    FALSE
## ENSG00000240361.1                       FALSE                    FALSE
```

```
# Summary of how many TRUEs there are in each row
# There are 13142 genes that have TRUEs in all 30 samples.
table(rowSums(thresh))
```

```
##
##      0      1      2      3      4      5      6      7      8      9     10     11     12
## 37003    825    439    332    294    244    211    174    164    180    140    154    125
##     13     14     15     16     17     18     19     20     21     22     23     24     25
##    130    139    136    142    125    124    121    113    138    108    117    148    161
##     26     27     28     29     30
##    174    206    256    437  13142
```

```
# we would like to keep genes that have at least 2 TRUES in each row of thresh
keep <- rowSums(thresh) >= 2
# Subset the rows of countdata to keep the more highly expressed genes
counts.keep <- scounts[keep,]
summary(keep)
```

```
##     Mode    FALSE    TRUE
## logical    37828   18374
```

```
dim(counts.keep)
```

```
## [1] 18374    30
```

En este caso, se usa un CPM de 0.5 ya que corresponde a un "recuento por gen" de 10-15 segun los "library size" de este conjunto de datos.

Asimismo se utiliza la condición de que la la expresión sea en 2 o más "libraries"" ya que en este caso cada situación experimental contiene dos replicas y ello nos "asegura" que "analizaremos" genes que como mínimo se expresen en un grupo.

Como regla general, se puede elegir un buen umbral identificando el CPM que corresponde a un recuento de 10.

Se debe filtrar a partir de el objeto CPM en lugar de filtrar los counting data (recuentos directamente), ya que este último no tiene en cuenta las diferencias en los tamaños de biblioteca (library sizes) entre las muestras.

```
# Let's have a look and see whether our threshold of 0.5 does indeed correspond to a
 count of about 10-15
# We will look at the first sample
plot(counts_cpm[,1],scounts[,1])
```
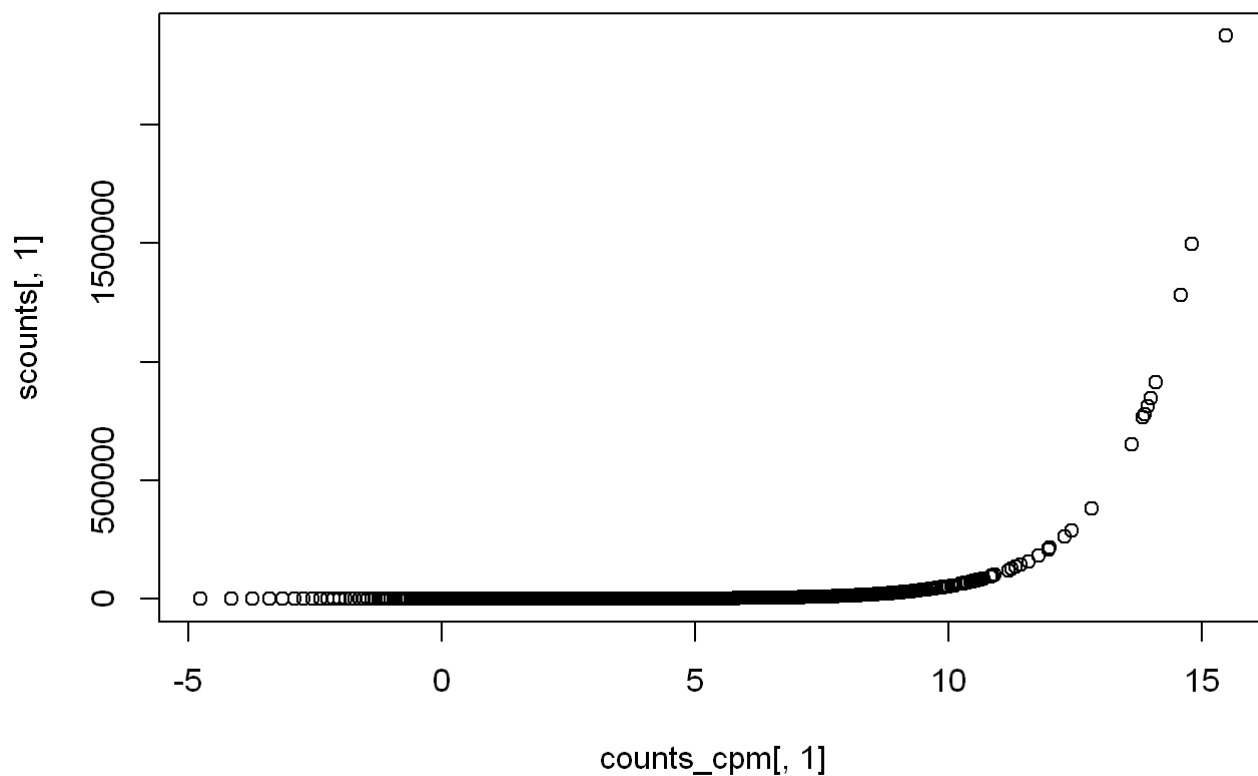
7. Counts to DGEList object

A continuación crearemos un objeto DGEList. Este es un objeto utilizado por edgeR para almacenar datos de recuento

```
y <- DGEList(counts.keep)
# have a look at y
y
```

```
## An object of class "DGEList"
## $counts
##                   GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000227232.4                      474                      669
## ENSG00000237683.5                      427                      791
## ENSG00000241860.2                       92                       68
## ENSG00000228463.4                        9                        9
## ENSG00000225972.1                       87                      155
##                   GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000227232.4                      786                      408
## ENSG00000237683.5                      553                      800
## ENSG00000241860.2                      107                       29
## ENSG00000228463.4                       57                       51
## ENSG00000225972.1                       82                      135
##                   GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000227232.4                     1301                      633
## ENSG00000237683.5                     1132                     1075
## ENSG00000241860.2                       46                       82
## ENSG00000228463.4                       92                       16
## ENSG00000225972.1                       23                       57
##                   GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000227232.4                      369                      430
## ENSG00000237683.5                      235                      679
## ENSG00000241860.2                       21                       95
## ENSG00000228463.4                        6                       26
## ENSG00000225972.1                      101                       54
##                   GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000227232.4                     1002                      460
## ENSG00000237683.5                      602                      279
## ENSG00000241860.2                       96                       41
## ENSG00000228463.4                       39                       16
## ENSG00000225972.1                     4678                       99
##                   GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000227232.4                      719                      825
## ENSG00000237683.5                     1064                      853
## ENSG00000241860.2                       91                       94
## ENSG00000228463.4                       41                       98
## ENSG00000225972.1                       64                       55
##                   GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000227232.4                      636                      775
## ENSG00000237683.5                      606                      580
## ENSG00000241860.2                       29                       69
## ENSG00000228463.4                       91                        5
## ENSG00000225972.1                       18                       37
##                   GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000227232.4                      834                      423
## ENSG00000237683.5                      445                      325
## ENSG00000241860.2                       40                       41
## ENSG00000228463.4                       66                       42
## ENSG00000225972.1                       33                       25
##                   GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
## ENSG00000227232.4                      297                      422
## ENSG00000237683.5                      229                      247
## ENSG00000241860.2                       50                       73
## ENSG00000228463.4                       53                       56
## ENSG00000225972.1                      192                       77
##                   GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
```

```
## ENSG00000227232.4                      541                      820
## ENSG00000237683.5                     1468                      973
## ENSG00000241860.2                       85                      101
## ENSG00000228463.4                       35                       66
## ENSG00000225972.1                       50                       49
##                    GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000227232.4                      766                      568
## ENSG00000237683.5                      374                      738
## ENSG00000241860.2                       71                       67
## ENSG00000228463.4                       28                       52
## ENSG00000225972.1                       74                       81
##                    GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000227232.4                      482                      365
## ENSG00000237683.5                      681                      359
## ENSG00000241860.2                       63                       61
## ENSG00000228463.4                       51                       15
## ENSG00000225972.1                      110                       82
##                    GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000227232.4                     1472                     1325
## ENSG00000237683.5                     2020                      853
## ENSG00000241860.2                      196                       94
## ENSG00000228463.4                       52                       44
## ENSG00000225972.1                       54                       50
##                    GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000227232.4                      450                      689
## ENSG00000237683.5                      352                      809
## ENSG00000241860.2                       43                       82
## ENSG00000228463.4                        9                       61
## ENSG00000225972.1                      116                       37
##                    GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000227232.4                      838                     1003
## ENSG00000237683.5                     1212                      960
## ENSG00000241860.2                       89                       59
## ENSG00000228463.4                       21                       26
## ENSG00000225972.1                       51                       66
## 18369 more rows ...
##
## $samples
##                           group lib.size norm.factors
## GTEX.111VG.0526.SM.5N9BW     1 52085501            1
## GTEX.11EM3.0126.SM.5985K     1 64954617            1
## GTEX.11EMC.0226.SM.5EGLP     1 65673287            1
## GTEX.11NSD.0126.SM.5987F     1 52084492            1
## GTEX.11NV4.0626.SM.5N9BR     1 51837308            1
## 25 more rows ...
```

```
# See what slots are stored in y
names(y)
```

```
## [1] "counts"  "samples"
```

```
# Library size information is stored in the samples slot
y$samples
```

```
##                           group lib.size norm.factors
## GTEX.111VG.0526.SM.5N9BW     1 52085501            1
## GTEX.11EM3.0126.SM.5985K     1 64954617            1
## GTEX.11EMC.0226.SM.5EGLP     1 65673287            1
## GTEX.11NSD.0126.SM.5987F     1 52084492            1
## GTEX.11NV4.0626.SM.5N9BR     1 51837308            1
## GTEX.11O72.2326.SM.5BC7H     1 68388840            1
## GTEX.12WSG.0226.SM.5EGIF     1 38526920            1
## GTEX.139UW.0126.SM.5KM1B     1 48121274            1
## GTEX.13NZ9.1126.SM.5MR37     1 61301417            1
## GTEX.13O1R.0826.SM.5J2MB     1 60635354            1
## GTEX.13OVG.0226.SM.5LU93     1 47021443            1
## GTEX.13QJC.0826.SM.5RQKC     1 48725791            1
## GTEX.13U4I.0526.SM.5LU59     1 45294109            1
## GTEX.14ABY.0926.SM.5Q5DY     1 64593320            1
## GTEX.14AS3.0226.SM.5Q5B6     1 41908407            1
## GTEX.14BMU.0226.SM.5S2QA     1 44406229            1
## GTEX.PWN1.2626.SM.2I3FH      1 39644072            1
## GTEX.S7SE.0726.SM.2XCD7      1 41928929            1
## GTEX.T5JW.1226.SM.3GACY      1 46317315            1
## GTEX.WYVS.0326.SM.3NM9V      1 76564730            1
## GTEX.XBED.0126.SM.47JY7      1 47027068            1
## GTEX.XMK1.0626.SM.4B65A      1 52878007            1
## GTEX.Y5V6.0526.SM.4VBRV      1 68606086            1
## GTEX.YEC4.0626.SM.5CVLU      1 50583432            1
## GTEX.YFC4.2626.SM.5P9FQ      1 80995956            1
## GTEX.YJ89.0726.SM.5P9F7      1 73817346            1
## GTEX.Z9EW.0226.SM.5CVM7      1 46741531            1
## GTEX.ZLV1.0126.SM.4WWBZ      1 55313069            1
## GTEX.ZYVF.1126.SM.5E458      1 46918338            1
## GTEX.ZYY3.1926.SM.5GZXS      1 48818257            1
```

8. Quality control

Ahora que hemos eliminado los genes de baja expresión y hemos almacenado nuestros conteos en un objeto DGEList, vamos a llevar a cabo algunos gráficos que nos permitan realizar un pequeño informe de los mismos (Quality control).

Library sizes and distribution plots

Primero, podemos verificar cuántas lecturas tenemos para cada muestra en el objeto creado (counting data)

```
y$samples$lib.size
```

```
##  [1] 52085501 64954617 65673287 52084492 51837308 68388840 38526920 48121274
##  [9] 61301417 60635354 47021443 48725791 45294109 64593320 41908407 44406229
## [17] 39644072 41928929 46317315 76564730 47027068 52878007 68606086 50583432
## [25] 80995956 73817346 46741531 55313069 46918338 48818257
```

9. Diagrama de barras de los library sizes

También podemos plotear a partir de un diagrama de barras de los "library sizes"" para ver si hay discrepancias importantes entre las muestras

```
# The names argument tells the barplot to use the sample names on the x-axis
# The las argument rotates the axis names
barplot(y$samples$lib.size,names=colnames(y),las=2)
# Add a title to the plot
title("Barplot of library sizes")
```

Barplot of library sizes
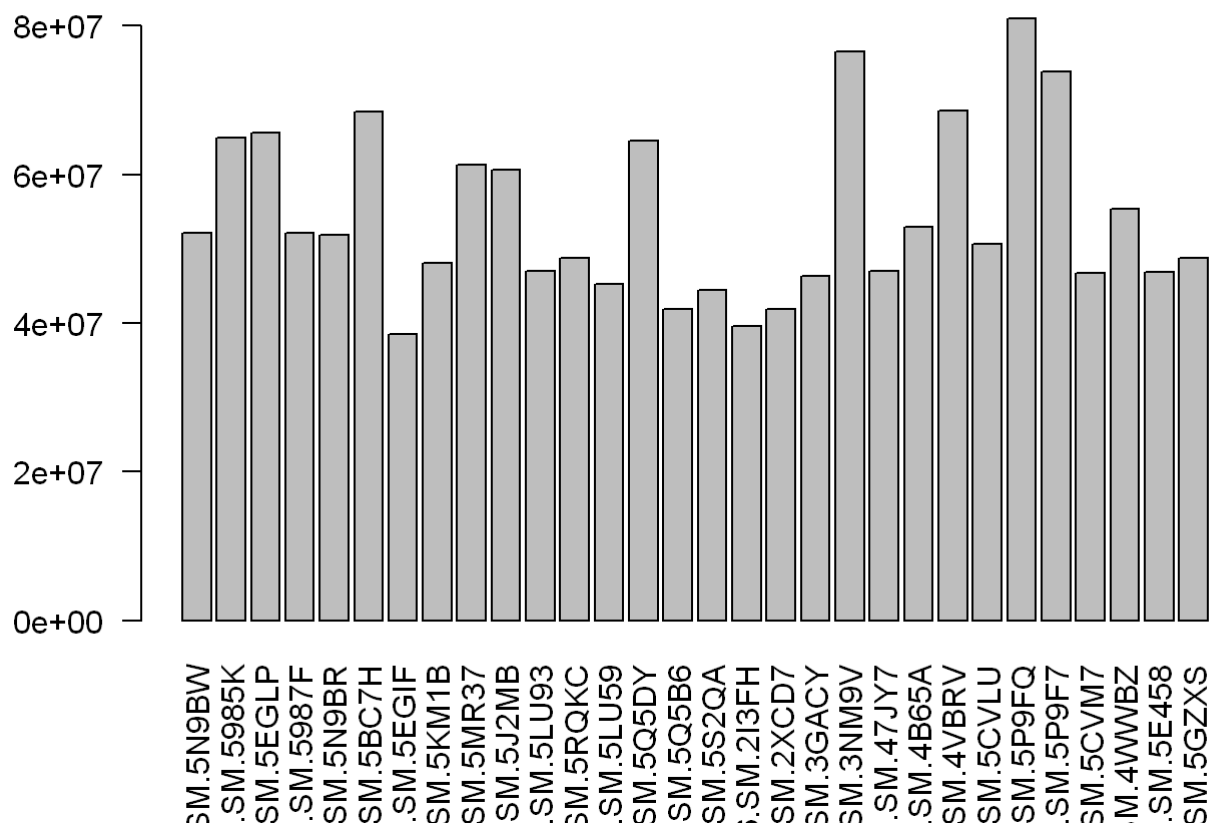
Los "ounting data" (datos de recuento) no se distribuyen segun una Distribución Normal, por lo que si queremos examinar las distribuciones de los recuentos sin procesar, utilizaremos Boxplots para verificar la distribución de los recuentos de lectura en escala log2.

Podemos usar la función cpm para obtener recuentos de log2 por millón, corregidos por los library sizes (tamaños de biblioteca). La función cpm también incorpora una pequeña "modificación" para evitar el problema asociado al logaritmo de valores de cero.

10, Diagrama de cajas

```
# Get log2 counts per million
logcounts <- cpm(y,log=TRUE)
# Check distributions of samples using boxplots
boxplot(logcounts, xlab="", ylab="Log2 counts per million",las=2)
# Let's add a blue horizontal line that corresponds to the median logCPM
abline(h=median(logcounts),col="blue")
title("Boxplots of logCPMs (unnormalised)")
```

Boxplots of logCPMs (unnormalised)
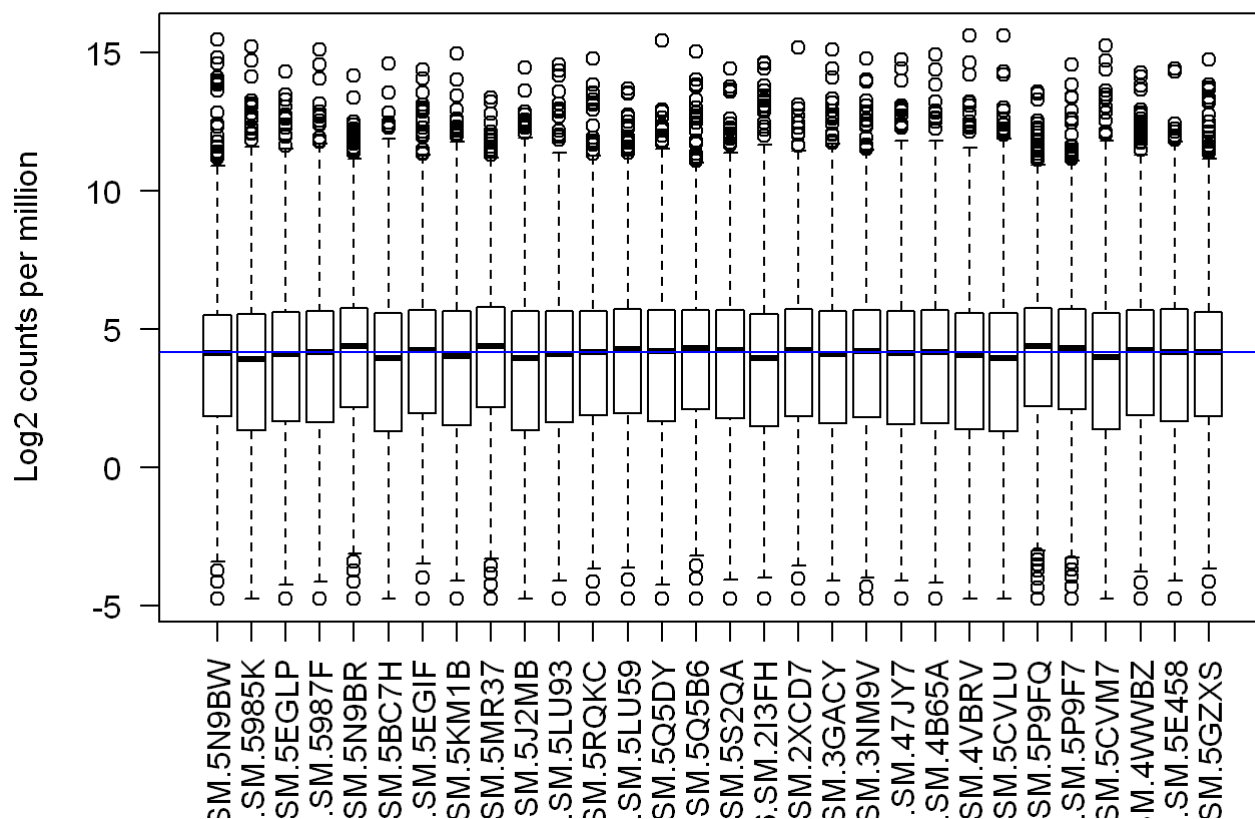
De los boxplots, vemos que, en general, las distribuciones del counting data no son idénticas, pero tampoco son muy diferentes.

Si una muestra está realmente muy por encima o por debajo de la línea horizontal azul, es posible que tengamos que investigar más esa muestra.

11. Multidimensional scaling plots

Uno de los gráficos más importante en el Quality control es el MDS. Un MDSplot es un gráfico, que nos permite "visualizar" variabilidad en los datos. Si su experimento está bien "controlado"" y funcionó bien, lo que esperamos ver es que las principales fuentes de variación en los datos sean los tratamientos / grupos que nos interesan.

También nos puede ayudar en "la visualización de valores atípicos. Podemos usar la función plotMDS para crear el diagrama de MDS.

Diagrama de MDS

```
plotMDS(y)
```

La verdad es que no se ve muy bien :(

12. Hierarchical clustering with heatmaps
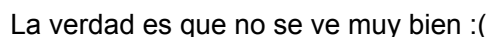
Podemos complementar la visualización de los datos con la función heapmap.2 que nos permite+iria obtener la representación del cluster jerárquico de las muestras, en concreto, en este ejemplo, se grafica (a partir del método average) la matriz de distancias euclídeas del logCPM (objeto logcounts) para los 500 genes más variables. El diagrama del heatmap se representará en el último apartado.

```
# We estimate the variance for each row in the logcounts matrix
var_genes <- apply(logcounts, 1, var)
head(var_genes)
```

```
## ENSG00000227232.4 ENSG00000237683.5 ENSG00000241860.2 ENSG00000228463.4
##         0.2566127         0.5135544         0.2989137         1.2509434
## ENSG00000225972.1 ENSG00000225630.1
##         1.8498334         0.7374741
```

```
# Get the gene names for the top 500 most variable genes
select_var <- names(sort(var_genes, decreasing=TRUE))[1:500]
head(select_var)
```

```
## [1] "ENSG00000229807.5"  "ENSG00000110680.8"  "ENSG00000012817.11"
## [4] "ENSG00000114374.8"  "ENSG00000131002.7"  "ENSG00000129824.11"
```

```r
# Subset logcounts matrix
highly_variable_lcpm <- logcounts[select_var,]
dim(highly_variable_lcpm)
```

```
## [1] 500  30
```

```r
head(highly_variable_lcpm)
```

```
##                      GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000229807.5                  -2.718176                9.3032207
## ENSG00000110680.8                  -1.622558               -0.5821209
## ENSG00000012817.11                  7.263680               -1.8206317
## ENSG00000114374.8                   6.060782               -2.1745202
## ENSG00000131002.7                   6.654994               -2.7904720
## ENSG00000129824.11                  7.462084               -0.8761533
##                      GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000229807.5                  8.8746914               -2.128050
## ENSG00000110680.8                  7.7763875               -2.718155
## ENSG00000012817.11                -2.8022915                7.275198
## ENSG00000114374.8                 -2.1877475                6.507863
## ENSG00000131002.7                 -2.6566599                6.091193
## ENSG00000129824.11                -0.8909509                7.677126
##                      GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000229807.5                  -2.712949               -1.602891
## ENSG00000110680.8                  -2.712949                3.397195
## ENSG00000012817.11                  7.871551                6.678998
## ENSG00000114374.8                   6.485204                6.132382
## ENSG00000131002.7                   7.409647                5.157281
## ENSG00000129824.11                  7.852049                7.846139
##                      GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000229807.5                  9.4114330               -2.300322
## ENSG00000110680.8                 -1.8869714               -1.805201
## ENSG00000012817.11                -3.9947683                6.975253
## ENSG00000114374.8                 -3.4951997                6.810346
## ENSG00000131002.7                 -3.1248326                6.436111
## ENSG00000129824.11                -0.6000913                7.553627
##                      GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000229807.5                  -3.848986               -1.384193
## ENSG00000110680.8                  -2.445467                9.692514
## ENSG00000012817.11                  7.977681                6.677879
## ENSG00000114374.8                   6.397054                6.536891
## ENSG00000131002.7                   6.683575                6.145760
## ENSG00000129824.11                  7.784346                7.762029
##                      GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000229807.5                  -2.003235                9.434566
## ENSG00000110680.8                  -3.036955               -1.299790
## ENSG00000012817.11                  7.366544               -3.072626
## ENSG00000114374.8                   6.597902               -4.125467
## ENSG00000131002.7                   6.512768               -3.683767
## ENSG00000129824.11                  7.557474               -1.820885
##                      GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000229807.5                  9.6226138               -1.1867559
## ENSG00000110680.8                  0.3231476                0.7197045
## ENSG00000012817.11                -2.9990437                6.5049404
## ENSG00000114374.8                 -4.0868357                6.3479110
## ENSG00000131002.7                 -3.2792143                5.8086103
## ENSG00000129824.11                -1.3583875                7.9679308
##                      GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000229807.5                  9.6504170                9.805710
## ENSG00000110680.8                 -4.7652806               -3.611581
## ENSG00000012817.11                -1.4311584               -2.978820
## ENSG00000114374.8                 -1.9912251               -4.765281
## ENSG00000131002.7                 -3.5650022               -3.611581
## ENSG00000129824.11                -0.6056577               -2.978820
##                      GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
```

```
## ENSG00000229807.5                  9.7351668                -3.2066175
## ENSG00000110680.8                  0.7668791                -0.7727064
## ENSG00000012817.11                -1.2683373                 6.9801434
## ENSG00000114374.8                 -2.8607245                 6.3623148
## ENSG00000131002.7                 -3.5191958                 6.1859211
## ENSG00000129824.11                -1.7907803                 8.1136462
##                     GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
## ENSG00000229807.5                  8.9684011                10.0114465
## ENSG00000110680.8                  6.2705314                 2.4060716
## ENSG00000012817.11                 0.2134548                -2.5788043
## ENSG00000114374.8                 -1.0936478                -1.5684495
## ENSG00000131002.7                 -0.4164854                -2.3695848
## ENSG00000129824.11                -0.2968670                -0.7726863
##                     GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000229807.5                 -1.776240                 -2.272477
## ENSG00000110680.8                 10.001647                  9.879521
## ENSG00000012817.11                 7.177538                  6.686635
## ENSG00000114374.8                  6.682904                  6.318754
## ENSG00000131002.7                  6.279340                  6.166787
## ENSG00000129824.11                 7.791057                  7.714708
##                     GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000229807.5                 -2.848890                 -0.967942
## ENSG00000110680.8                 10.194599                  1.721901
## ENSG00000012817.11                 6.711817                  7.010032
## ENSG00000114374.8                  6.037534                  6.076172
## ENSG00000131002.7                  5.757531                  5.865838
## ENSG00000129824.11                 7.404991                  7.569556
##                     GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000229807.5                  9.374923                 -1.413327
## ENSG00000110680.8                 -3.536899                 -3.082495
## ENSG00000012817.11                -2.757425                  7.297902
## ENSG00000114374.8                 -4.765281                  6.496223
## ENSG00000131002.7                 -4.765281                  6.638566
## ENSG00000129824.11                -2.641747                  7.947008
##                     GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000229807.5                 -0.1640095                 9.459529
## ENSG00000110680.8                  9.9851052                -2.085349
## ENSG00000012817.11                 7.2177655                -4.765281
## ENSG00000114374.8                  6.2123057                -2.975218
## ENSG00000131002.7                  5.9585426                -3.777364
## ENSG00000129824.11                 7.5016355                -3.196451
##                     GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000229807.5                  9.848223161                8.5286664
## ENSG00000110680.8                 -2.270345829                3.7565253
## ENSG00000012817.11                -0.983729159               -1.1634167
## ENSG00000114374.8                 -3.654745620               -2.4726482
## ENSG00000131002.7                 -3.311691623               -2.4726482
## ENSG00000129824.11                -0.005949852               -0.7615973
```

13. Normalization for "composition bias"

El procesos de normalización denominado TMM se realiza para eliminar los sesgos de composición (bias compostion) entre las bibliotecas.

Este método genera un conjunto de factores de normalización, donde el producto de estos factores y los tamaños de la biblioteca (library sizes) definen el tamaño efectivo de la biblioteca (effective library size).

La función calcNormFactors calcula los factores de normalización entre bibliotecas.

```
# Apply normalisation to DGEList object
y <- calcNormFactors(y)
head(y)
```

```
## An object of class "DGEList"
## $counts
##                      GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000227232.4                         474                      669
## ENSG00000237683.5                         427                      791
## ENSG00000241860.2                          92                       68
## ENSG00000228463.4                           9                        9
## ENSG00000225972.1                          87                      155
## ENSG00000225630.1                       31646                    12906
##                      GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000227232.4                         786                      408
## ENSG00000237683.5                         553                      800
## ENSG00000241860.2                         107                       29
## ENSG00000228463.4                          57                       51
## ENSG00000225972.1                          82                      135
## ENSG00000225630.1                        9595                     9332
##                      GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000227232.4                        1301                      633
## ENSG00000237683.5                        1132                     1075
## ENSG00000241860.2                          46                       82
## ENSG00000228463.4                          92                       16
## ENSG00000225972.1                          23                       57
## ENSG00000225630.1                        8718                    19638
##                      GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000227232.4                         369                      430
## ENSG00000237683.5                         235                      679
## ENSG00000241860.2                          21                       95
## ENSG00000228463.4                           6                       26
## ENSG00000225972.1                         101                       54
## ENSG00000225630.1                        8997                     7455
##                      GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000227232.4                        1002                      460
## ENSG00000237683.5                         602                      279
## ENSG00000241860.2                          96                       41
## ENSG00000228463.4                          39                       16
## ENSG00000225972.1                        4678                       99
## ENSG00000225630.1                        8450                    13275
##                      GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000227232.4                         719                      825
## ENSG00000237683.5                        1064                      853
## ENSG00000241860.2                          91                       94
## ENSG00000228463.4                          41                       98
## ENSG00000225972.1                          64                       55
## ENSG00000225630.1                       14271                    18090
##                      GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000227232.4                         636                      775
## ENSG00000237683.5                         606                      580
## ENSG00000241860.2                          29                       69
## ENSG00000228463.4                          91                        5
## ENSG00000225972.1                          18                       37
## ENSG00000225630.1                        7910                    10398
##                      GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000227232.4                         834                      423
## ENSG00000237683.5                         445                      325
## ENSG00000241860.2                          40                       41
## ENSG00000228463.4                          66                       42
## ENSG00000225972.1                          33                       25
```

```
## ENSG00000225630.1                        9934                    7614
##                      GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
## ENSG00000227232.4                         297                     422
## ENSG00000237683.5                         229                     247
## ENSG00000241860.2                          50                      73
## ENSG00000228463.4                          53                      56
## ENSG00000225972.1                         192                      77
## ENSG00000225630.1                       11276                    9076
##                      GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
## ENSG00000227232.4                         541                     820
## ENSG00000237683.5                        1468                     973
## ENSG00000241860.2                          85                     101
## ENSG00000228463.4                          35                      66
## ENSG00000225972.1                          50                      49
## ENSG00000225630.1                       12188                   14976
##                      GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000227232.4                         766                     568
## ENSG00000237683.5                         374                     738
## ENSG00000241860.2                          71                      67
## ENSG00000228463.4                          28                      52
## ENSG00000225972.1                          74                      81
## ENSG00000225630.1                       10542                   34546
##                      GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000227232.4                         482                     365
## ENSG00000237683.5                         681                     359
## ENSG00000241860.2                          63                      61
## ENSG00000228463.4                          51                      15
## ENSG00000225972.1                         110                      82
## ENSG00000225630.1                       14469                   36579
##                      GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000227232.4                        1472                    1325
## ENSG00000237683.5                        2020                     853
## ENSG00000241860.2                         196                      94
## ENSG00000228463.4                          52                      44
## ENSG00000225972.1                          54                      50
## ENSG00000225630.1                       12782                   27107
##                      GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000227232.4                         450                     689
## ENSG00000237683.5                         352                     809
## ENSG00000241860.2                          43                      82
## ENSG00000228463.4                           9                      61
## ENSG00000225972.1                         116                      37
## ENSG00000225630.1                       51193                    6712
##                      GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000227232.4                         838                    1003
## ENSG00000237683.5                        1212                     960
## ENSG00000241860.2                          89                      59
## ENSG00000228463.4                          21                      26
## ENSG00000225972.1                          51                      66
## ENSG00000225630.1                        6015                   48449
##
## $samples
##                           group lib.size norm.factors
## GTEX.111VG.0526.SM.5N9BW      1 52085501    0.9417468
## GTEX.11EM3.0126.SM.5985K      1 64954617    0.8841413
## GTEX.11EMC.0226.SM.5EGLP      1 65673287    0.9728161
## GTEX.11NSD.0126.SM.5987F      1 52084492    1.0252789
```

```
## GTEX.11NV4.0626.SM.5N9BR      1 51837308      1.1079924
## 25 more rows ...
```

Esta linea "actualizará"" los factores de normalización en el objeto DGEList (sus valores predeterminados son 1).

```
y$samples
```

```
##                          group lib.size norm.factors
## GTEX.111VG.0526.SM.5N9BW     1 52085501    0.9417468
## GTEX.11EM3.0126.SM.5985K     1 64954617    0.8841413
## GTEX.11EMC.0226.SM.5EGLP     1 65673287    0.9728161
## GTEX.11NSD.0126.SM.5987F     1 52084492    1.0252789
## GTEX.11NV4.0626.SM.5N9BR     1 51837308    1.1079924
## GTEX.11O72.2326.SM.5BC7H     1 68388840    0.9028173
## GTEX.12WSG.0226.SM.5EGIF     1 38526920    1.0468427
## GTEX.139UW.0126.SM.5KM1B     1 48121274    0.9521874
## GTEX.13NZ9.1126.SM.5MR37     1 61301417    1.1459429
## GTEX.13O1R.0826.SM.5J2MB     1 60635354    0.9216591
## GTEX.13OVG.0226.SM.5LU93     1 47021443    0.9875460
## GTEX.13QJC.0826.SM.5RQKC     1 48725791    0.9969542
## GTEX.13U4I.0526.SM.5LU59     1 45294109    1.0698486
## GTEX.14ABY.0926.SM.5Q5DY     1 64593320    1.0162815
## GTEX.14AS3.0226.SM.5Q5B6     1 41908407    1.0875705
## GTEX.14BMU.0226.SM.5S2QA     1 44406229    1.0165402
## GTEX.PWN1.2626.SM.2I3FH      1 39644072    0.9110875
## GTEX.S7SE.0726.SM.2XCD7      1 41928929    1.0824931
## GTEX.T5JW.1226.SM.3GACY      1 46317315    0.9887829
## GTEX.WYVS.0326.SM.3NM9V      1 76564730    1.0296372
## GTEX.XBED.0126.SM.47JY7      1 47027068    0.9998168
## GTEX.XMK1.0626.SM.4B65A      1 52878007    1.0106043
## GTEX.Y5V6.0526.SM.4VBRV      1 68606086    0.9260320
## GTEX.YEC4.0626.SM.5CVLU      1 50583432    0.9171709
## GTEX.YFC4.2626.SM.5P9FQ      1 80995956    1.1087606
## GTEX.YJ89.0726.SM.5P9F7      1 73817346    1.1032385
## GTEX.Z9EW.0226.SM.5CVM7      1 46741531    0.9145234
## GTEX.ZLV1.0126.SM.4WWBZ      1 55313069    0.9860888
## GTEX.ZYVF.1126.SM.5E458      1 46918338    1.0280662
## GTEX.ZYY3.1926.SM.5GZXS      1 48818257    0.9886884
```

Un factor de normalización por debajo de uno indica que el tamaño de la biblioteca se reducirá, ya que "hay más sesgo de composición" (composition bias) en esa biblioteca en relación con las otras bibliotecas.

Es decir estamos re-escalando los recuentos "incrementandolos" en esa muestra. Por el contrario, un factor por encima de uno es equivalente a "reesscalar a la baja" los recuentos.

Si graficamos la diferencias medias usando la función plotMD para estas muestras, deberíamos poder ver el problema de sesgo de composición (bias composition).

Utilizaremos los logcounts, "normalizados por el tamaño de la biblioteca" (library size)", pero no para el sesgo de composición (bias composition)

```
par(mfrow=c(1,2))
plotMD(logcounts,column = 7)
abline(h=0,col="grey")
plotMD(logcounts,column = 11)
abline(h=0,col="grey")
```

GTEX.12WSG.0226.SM.5EGIF   GTEX.13OVG.0226.SM.5LU93
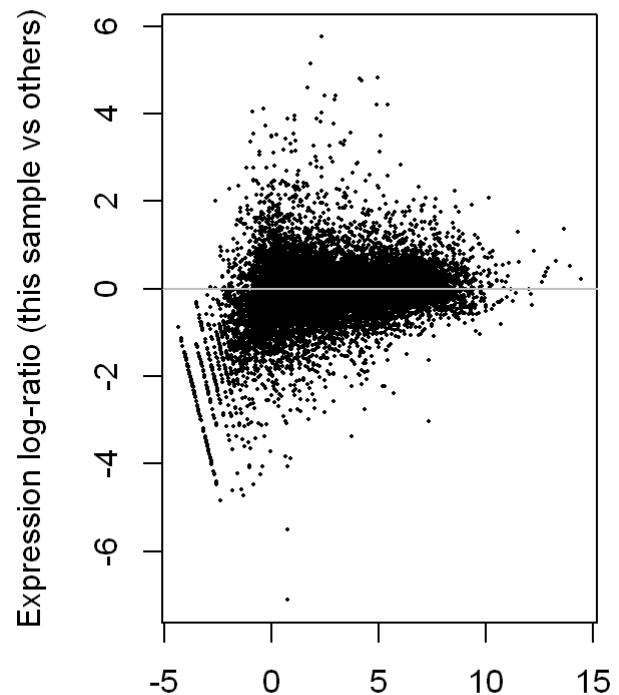
Los gráficos de "diferencia de medias"" muestran la expresión promedio (media: eje x) contra los cambios logfold (diferencia: eje y).

Debido a que nuestro objeto DGEList contiene los factores de normalización, si rehacemos estos gráficos usando y(el objeto y), deberíamos ver que el problema de sesgo de composición (bias composition) ha sido resuelto.

```
par(mfrow=c(1,2))
plotMD(y,column = 7)
abline(h=0,col="grey")
plotMD(y,column = 11)
abline(h=0,col="grey")
```

GTEX.12WSG.0226.SM.5EGIF
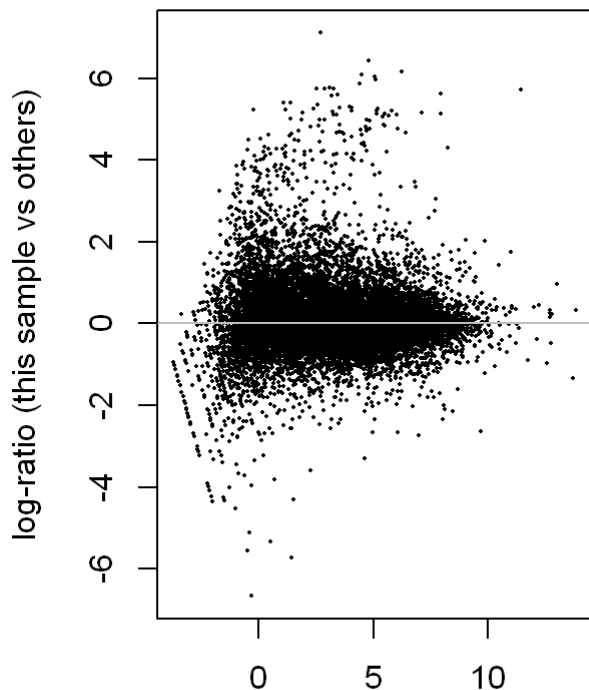
GTEX.13OVG.0226.SM.5LU93
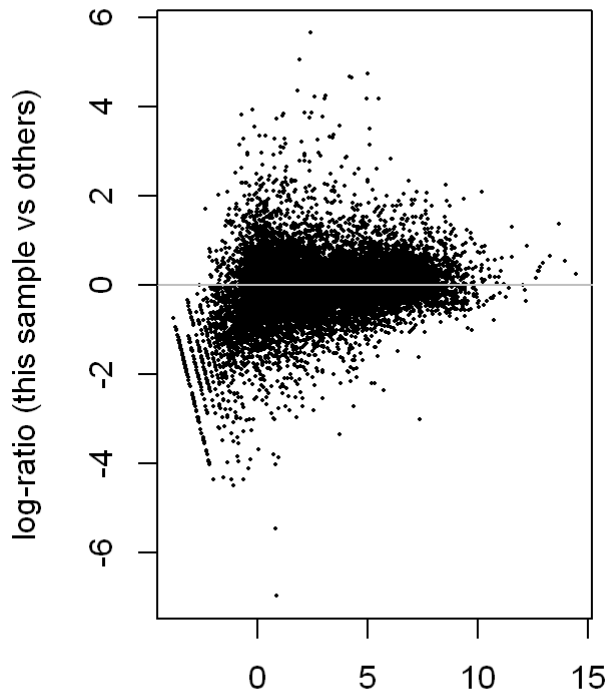
Average log CPM (this sample and others)   Average log CPM (this sample and others)

14. Differential expression with limma-voom

Hay una serie de paquetes para analizar datos de RNA-Seq. El paquete limma (Ritchie et al., 2015) (desde la versión 3.16.0) ofrece la función voom, que transforma los recuentos de lectura en logCPM teniendo en cuenta la relación de la media y varianza de los datos (Law et al., 2014).

Después de aplicar voom, los usuarios pueden aplicar un modelo lineal a los datos transformados por voom para identificar genes expresados diferencialmente, utilizando comandos estándar de limma.

Leemos los targets seleccionados y los guardamos en una variable stargets. Despues determinamos los factores y niveles que tenemos. Vemos que nos salen los 3 niveles: SFI, NIT y ELI

```
stargets<- read.csv2(file.path("./data", "stargets.csv"), head=T, sep=";")
head(stargets,5)
```

```
##   Experiment SRA_Sample           Sample_Name Grupo_analisis body_site
## 1  SRX223301  SRS389914  GTEX-T5JW-1226-SM-3GACY              2   Thyroid
## 2  SRX605452  SRS639261 GTEX-12WSG-0226-SM-5EGIF              2   Thyroid
## 3  SRX559198  SRS624031 GTEX-11O72-2326-SM-5BC7H              2   Thyroid
## 4  SRX614680  SRS644012 GTEX-13O1R-0826-SM-5J2MB              2   Thyroid
## 5  SRX597594  SRS637292  GTEX-ZLV1-0126-SM-4WWBZ              2   Thyroid
##          molecular_data_type    sex Group ShortName
## 1 Allele-Specific Expression female   SFI T5JW-_SFI
## 2              RNA Seq (NGS) female   SFI 12WSG_SFI
## 3              RNA Seq (NGS)   male   SFI 11O72_SFI
## 4              RNA Seq (NGS)   male   SFI 13O1R_SFI
## 5 Allele-Specific Expression female   SFI ZLV1-_SFI
```

```
group<-factor(stargets$Group)
group
```

```
##  [1] SFI SFI SFI SFI SFI SFI SFI SFI SFI SFI NIT NIT NIT NIT NIT NIT NIT NIT NIT
## [20] NIT ELI ELI ELI ELI ELI ELI ELI ELI ELI ELI
## Levels: ELI NIT SFI
```

Create the design matrix

Primero, necesitamos crear una matriz de diseño para los grupos (lo teneis como material de consulta la guía del usuario de limma para obtener más información sobre las matrices de diseño y ya fue trabajado en la primera parte del curso).

Hay muchas formas diferentes de configurar la matriz de diseño, y estan supeditadas a las comparaciones que se "quieren testar". En este análisis, supongamos que queremos testar las diferencias de estado (status) en los diferentes tipos por separado.

Por ejemplo, queremos saber qué genes se expresan diferencialmente

Anteriormente "hemos codificado como variable grupo", que lleva implicito "cell type and status".

Codificar de esta manera nos permite ser flexibles al especificar las comparaciones que nos interesan

```
# Look at group variable again
group
```

```
##  [1] SFI SFI SFI SFI SFI SFI SFI SFI SFI SFI NIT NIT NIT NIT NIT NIT NIT NIT NIT
## [20] NIT ELI ELI ELI ELI ELI ELI ELI ELI ELI ELI
## Levels: ELI NIT SFI
```

```
# Specify a design matrix without an intercept term
design <- model.matrix(~ 0 + group)
design
```

```
##     groupELI groupNIT groupSFI
## 1          0        0        1
## 2          0        0        1
## 3          0        0        1
## 4          0        0        1
## 5          0        0        1
## 6          0        0        1
## 7          0        0        1
## 8          0        0        1
## 9          0        0        1
## 10         0        0        1
## 11         0        1        0
## 12         0        1        0
## 13         0        1        0
## 14         0        1        0
## 15         0        1        0
## 16         0        1        0
## 17         0        1        0
## 18         0        1        0
## 19         0        1        0
## 20         0        1        0
## 21         1        0        0
## 22         1        0        0
## 23         1        0        0
## 24         1        0        0
## 25         1        0        0
## 26         1        0        0
## 27         1        0        0
## 28         1        0        0
## 29         1        0        0
## 30         1        0        0
## attr(,"assign")
## [1] 1 1 1
## attr(,"contrasts")
## attr(,"contrasts")$group
## [1] "contr.treatment"
```

```
## Make the column names of the design matrix a bit nicer
colnames(design) <- levels(group)
design
```

```
##      ELI NIT SFI
## 1     0   0   1
## 2     0   0   1
## 3     0   0   1
## 4     0   0   1
## 5     0   0   1
## 6     0   0   1
## 7     0   0   1
## 8     0   0   1
## 9     0   0   1
## 10    0   0   1
## 11    0   1   0
## 12    0   1   0
## 13    0   1   0
## 14    0   1   0
## 15    0   1   0
## 16    0   1   0
## 17    0   1   0
## 18    0   1   0
## 19    0   1   0
## 20    0   1   0
## 21    1   0   0
## 22    1   0   0
## 23    1   0   0
## 24    1   0   0
## 25    1   0   0
## 26    1   0   0
## 27    1   0   0
## 28    1   0   0
## 29    1   0   0
## 30    1   0   0
## attr(,"assign")
## [1] 1 1 1
## attr(,"contrasts")
## attr(,"contrasts")$group
## [1] "contr.treatment"
```

Cada columna de la matriz de diseño nos remite a las muestras que corresponden a cada grupo

voom estima la tendencia de la varianza respecto a la media en el counting data, para luego asignar un peso a cada observación en función de la predicción de la varianza (segun el modelo que nos da la tendencia). Los pesos se usan luego en el proceso de modelado lineal para ajustar la heterocedasticidad.

Asi pués voom ajustará automáticamente los tamaños de biblioteca (library size) utilizando norm.factors ya calculados.

La transformación de voom usa la matriz de diseño de experimento y produce un objeto EList.

Podemos agregar plot = TRUE para generar un gráfico de la tendencia de media-varianza.

Este diagrama es importante ya que nos "informa" de si hay algún gen con "alta variabilidad" en nuestros datos, y sobretodo porque nos indica si hemos filtrado los recuentos bajos adecuadamente.

Los recuentos log2 normalizados que nos aporta voom se pueden encontrar en v$E.

```
par(mfrow=c(1,1))
v <- voom(y,design,plot = TRUE)
```

voom: Mean-variance trend

log2(count size + 0.5)
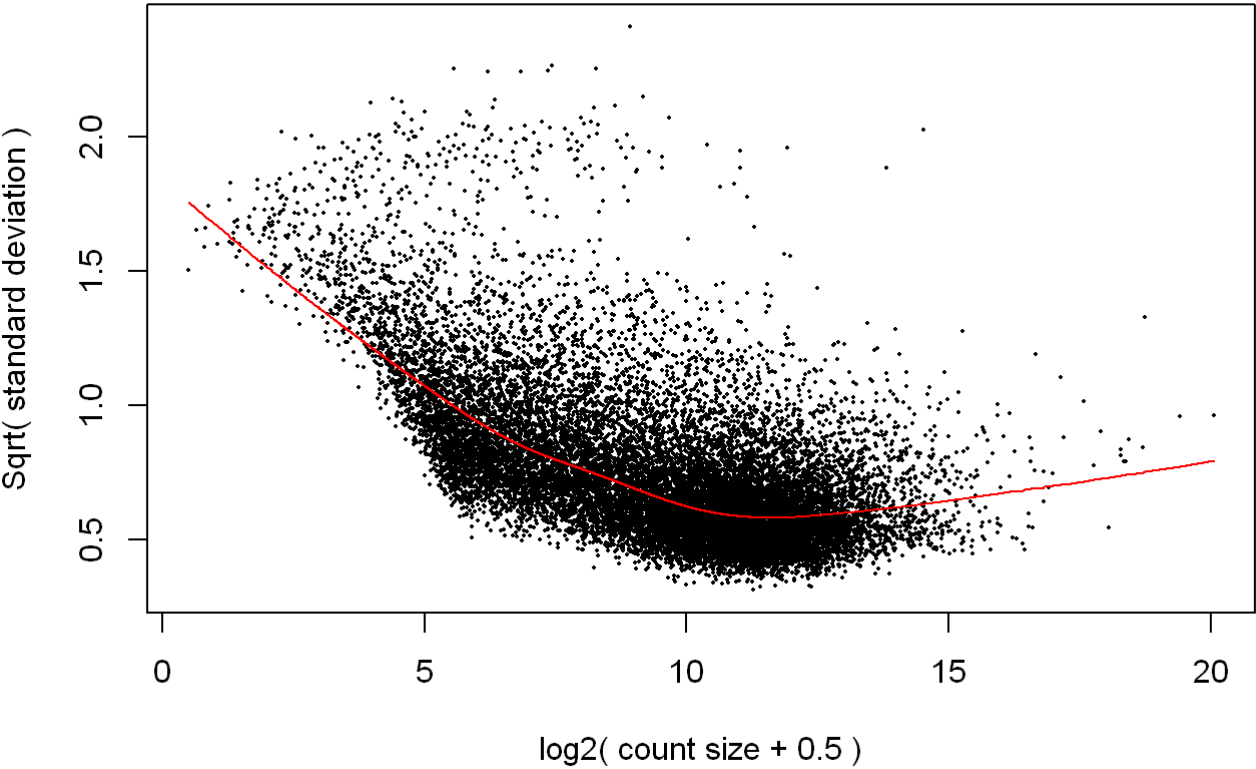
```
## An object of class "EList"
## $targets
##                          group lib.size norm.factors
## GTEX.111VG.0526.SM.5N9BW     1 49051354    0.9417468
## GTEX.11EM3.0126.SM.5985K     1 57429059    0.8841413
## GTEX.11EMC.0226.SM.5EGLP     1 63888030    0.9728161
## GTEX.11NSD.0126.SM.5987F     1 53401129    1.0252789
## GTEX.11NV4.0626.SM.5N9BR     1 57435341    1.1079924
## 25 more rows ...
##
## $E
##                  GTEX.111VG.0526.SM.5N9BW GTEX.11EM3.0126.SM.5985K
## ENSG00000227232.4                3.2740432                 3.543231
## ENSG00000237683.5                3.1235595                 3.784737
## ENSG00000241860.2                0.9151604                 0.254323
## ENSG00000228463.4               -2.3682936                -2.595782
## ENSG00000225972.1                0.8349900                 1.437062
##                  GTEX.11EMC.0226.SM.5EGLP GTEX.11NSD.0126.SM.5987F
## ENSG00000227232.4                3.6218292               2.93539391
## ENSG00000237683.5                3.1149657               3.90595924
## ENSG00000241860.2                0.7507191              -0.85615531
## ENSG00000228463.4               -0.1519837              -0.05229783
## ENSG00000225972.1                0.3688484               1.34335069
##                  GTEX.11NV4.0626.SM.5N9BR GTEX.11O72.2326.SM.5BC7H
## ENSG00000227232.4                4.5020927                3.3590057
## ENSG00000237683.5                4.3014285                4.1225967
## ENSG00000241860.2               -0.3047080                0.4181272
## ENSG00000228463.4                0.6875146               -1.9038009
## ENSG00000225972.1               -1.2892780               -0.1027050
##                  GTEX.12WSG.0226.SM.5EGIF GTEX.139UW.0126.SM.5KM1B
## ENSG00000227232.4                3.1955909                3.2319490
## ENSG00000237683.5                2.5457436                3.8904094
## ENSG00000241860.2               -0.9075749                1.0595084
## ENSG00000228463.4               -2.6333999               -0.7899999
## ENSG00000225972.1                1.3314963                0.2502639
##                  GTEX.13NZ9.1126.SM.5MR37 GTEX.13O1R.0826.SM.5J2MB
## ENSG00000227232.4                3.8350028                3.0426648
## ENSG00000237683.5                3.1004337                2.3223120
## ENSG00000241860.2                0.4580733               -0.4293531
## ENSG00000228463.4               -0.8306030               -1.7599984
## ENSG00000225972.1                6.0574466                0.8322321
##                  GTEX.13OVG.0226.SM.5LU93 GTEX.13QJC.0826.SM.5RQKC
## ENSG00000227232.4                3.9536841                4.0869116
## ENSG00000237683.5                4.5187934                4.1350345
## ENSG00000241860.2                0.9785330                0.9600296
## ENSG00000228463.4               -0.1621274                1.0198390
## ENSG00000225972.1                0.4740604                0.1922030
##                  GTEX.13U4I.0526.SM.5LU59 GTEX.14ABY.0926.SM.5Q5DY
## ENSG00000227232.4                3.7153585               3.56236978
## ENSG00000237683.5                3.6457056               3.14453906
## ENSG00000241860.2               -0.7160152               0.08232788
## ENSG00000228463.4                0.9170416              -3.57718158
## ENSG00000225972.1               -1.3892049              -0.80779450
##                  GTEX.14AS3.0226.SM.5Q5B6 GTEX.14BMU.0226.SM.5S2QA
## ENSG00000227232.4                4.1944915               3.22986068
## ENSG00000237683.5                3.2890049               2.85015626
## ENSG00000241860.2               -0.1704267              -0.12131805
```

```
## ENSG00000228463.4              0.5450057             -0.08696654
## ENSG00000225972.1             -0.4441876             -0.82393214
##                  GTEX.PWN1.2626.SM.2I3FH GTEX.S7SE.0726.SM.2XCD7
## ENSG00000227232.4              3.0420511              3.2185756
## ENSG00000237683.5              2.6676556              2.4470528
## ENSG00000241860.2              0.4835167              0.6954404
## ENSG00000228463.4              0.5667723              0.3159470
## ENSG00000225972.1              2.4140199              0.7718925
##                  GTEX.T5JW.1226.SM.3GACY GTEX.WYVS.0326.SM.3NM9V
## ENSG00000227232.4              3.5636121              3.3796154
## ENSG00000237683.5              5.0029221              3.6262930
## ENSG00000241860.2              0.9006470              0.3645917
## ENSG00000228463.4             -0.3674584             -0.2454617
## ENSG00000225972.1              0.1410060             -0.6713875
##                  GTEX.XBED.0126.SM.47JY7 GTEX.XMK1.0626.SM.4B65A
## ENSG00000227232.4              4.0269869              3.41120241
## ENSG00000237683.5              2.9936668              3.78863998
## ENSG00000241860.2              0.6047162              0.33700147
## ENSG00000228463.4             -0.7222651             -0.02556861
## ENSG00000225972.1              0.6640134              0.60891402
##                  GTEX.Y5V6.0526.SM.4VBRV GTEX.YEC4.0626.SM.5CVLU
## ENSG00000227232.4              2.9249865268              2.9778721
## ENSG00000237683.5              3.4231712414              2.9539925
## ENSG00000241860.2             -0.0007139186              0.4066590
## ENSG00000228463.4             -0.3028980782             -1.5816591
## ENSG00000225972.1              0.7985039540              0.8304668
##                  GTEX.YFC4.2626.SM.5P9FQ GTEX.YJ89.0726.SM.5P9F7
## ENSG00000227232.4              4.0353260              4.0246882
## ENSG00000237683.5              4.4917708              3.3896146
## ENSG00000241860.2              1.1296596              0.2146097
## ENSG00000228463.4             -0.7744804             -0.8718993
## ENSG00000225972.1             -0.7205416             -0.6894212
##                  GTEX.Z9EW.0226.SM.5CVM7 GTEX.ZLV1.0126.SM.4WWBZ
## ENSG00000227232.4              3.39765822              3.6600687
## ENSG00000237683.5              3.04375438              3.8915493
## ENSG00000241860.2              0.02521842              0.5969842
## ENSG00000228463.4             -2.16979756              0.1731765
## ENSG00000225972.1              1.44646107             -0.5405193
##                  GTEX.ZYVF.1126.SM.5E458 GTEX.ZYY3.1926.SM.5GZXS
## ENSG00000227232.4              4.11965373              4.3778882
## ENSG00000237683.5              4.65175578              4.3147051
## ENSG00000241860.2              0.89180253              0.3018811
## ENSG00000228463.4             -1.16574849             -0.8650162
## ENSG00000225972.1              0.09448728              0.4623458
## 18369 more rows ...
##
## $weights
##            [,1]      [,2]     [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 4.5030196 4.9672876 5.305353 4.7482229 4.967624 5.1961140 3.9941982
## [2,] 4.4399822 4.8979258 5.233247 4.6808738 4.898256 5.1227426 3.9400419
## [3,] 1.1203512 1.2607530 1.363865 1.1944368 1.260855 1.3303272 0.9653701
## [4,] 0.5522524 0.6176995 0.667198 0.5864319 0.617748 0.6508351 0.4819840
## [5,] 1.8087636 1.9948351 2.124401 1.9084084 1.994965 2.0829684 1.5866789
##            [,8]      [,9]     [,10]     [,11]     [,12]     [,13]     [,14]
## [1,] 4.3171825 5.6181665 4.8844544 4.6240495 4.7564756 4.7491173 5.717847
## [2,] 4.2570748 5.5410157 4.8164980 4.5022091 4.6302729 4.6231577 5.573086
## [3,] 1.0638239 1.4604738 1.2357549 1.1996647 1.2406274 1.2383870 1.542690
## [4,] 0.5264915 0.7152142 0.6057789 0.9451511 0.9783536 0.9765018 1.228841
```

```
## [5,] 1.7297023 2.2389950 1.9628821 0.9931146 1.0274919 1.0255752 1.288662
##             [,15]      [,16]      [,17]      [,18]      [,19]      [,20]      [,21]
## [1,] 4.5706526 4.5433151 3.9634813 4.5587553 4.5843372 6.325885 4.8684665
## [2,] 4.4505626 4.4241039 3.8631540 4.4390547 4.4637991 6.186931 4.8656209
## [3,] 1.1828085 1.1741993 0.9918591 1.1790600 1.1871233 1.747588 1.2779043
## [4,] 0.9318373 0.9250376 0.7816860 0.9288767 0.9352453 1.405188 0.6561726
## [5,] 0.9789747 0.9717428 0.8205094 0.9758258 0.9825997 1.472476 1.1794696
##             [,22]      [,23]      [,24]      [,25]      [,26]      [,27]      [,28]
## [1,] 5.2699987 5.8394604 4.8286008 6.981922 6.6647792 4.587306 5.3350296
## [2,] 5.2670029 5.8363192 4.8257843 6.978751 6.6617867 4.584592 5.3319875
## [3,] 1.4033024 1.5866558 1.2653449 1.986180 1.8696105 1.190606 1.4241754
## [4,] 0.7206152 0.8198366 0.6498234 1.066296 0.9900518 0.612590 0.7316222
## [5,] 1.2973029 1.4713290 1.1675956 1.858008 1.7448538 1.098021 1.3171905
##             [,29]      [,30]
## [1,] 4.9457510 4.9477069
## [2,] 4.9428488 4.9448032
## [3,] 1.3023259 1.3029452
## [4,] 0.6684714 0.6687823
## [5,] 1.2025686 1.2031545
## 18369 more rows ...
##
## $design
##   ELI NIT SFI
## 1   0   0   1
## 2   0   0   1
## 3   0   0   1
## 4   0   0   1
## 5   0   0   1
## 25 more rows ...
```

Ahora podemos comparar los boxplot despues antes y despues de normalizar. Los valores de expresión en v$E ya son valores en escala logarítmica log2.

```
par(mfrow=c(1,2))
boxplot(logcounts, xlab="", ylab="Log2 counts per million",las=2,main="Non normalised
logCPM")
## Let's add a blue horizontal line that corresponds to the median logCPM
abline(h=median(logcounts),col="blue")
boxplot(v$E, xlab="", ylab="Log2 counts per million",las=2,main="Voom transformed log
CPM")
## Let's add a blue horizontal line that corresponds to the median logCPM
abline(h=median(v$E),col="blue")
```

Non normalised logCPM
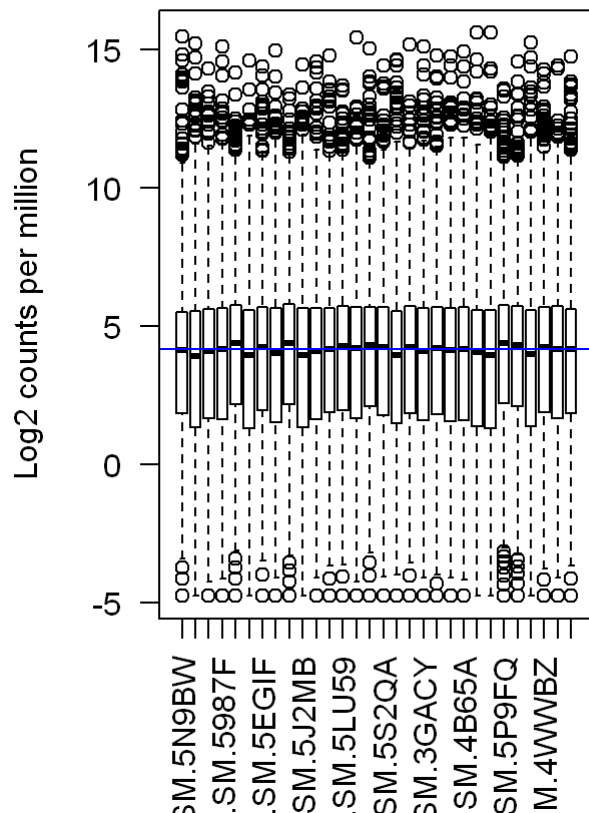
Voom transformed logCPM
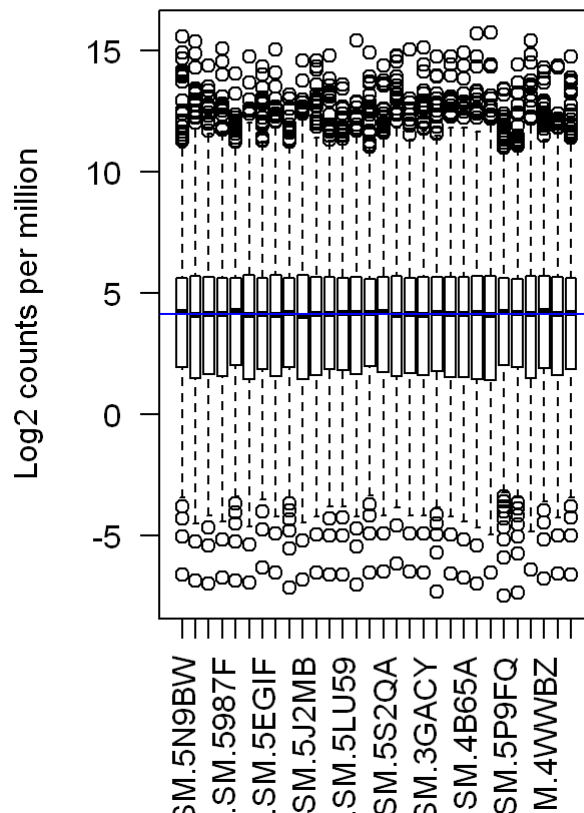
15. Testing for differential expression

Ahora que tenemos los datos obtenidos a partir de la función voom, podemos usar limma para obtener la expresión diferencial. Primero ajustamos un modelo lineal para cada gen usando la función lmFit en limma. lmFit necesita el objeto voom y la matriz de diseño que ya hemos especificado, que se encuentra dentro del objeto generado por voom

```
# Fit the linear model
fit <- lmFit(v)
names(fit)
```

```
##  [1] "coefficients"      "stdev.unscaled"  "sigma"           "df.residual"
##  [5] "cov.coefficients"  "pivot"           "rank"            "Amean"
##  [9] "method"            "design"
```

Hay una serie de elementos dentro del objetofit la mayoría de los cuales, son prácticamente idénticos a los vistos cuando aplicamos dicha función en la primera parte del microarray data analysis.

Dado que estamos interesados♂en obtener genes diferencialment expresados entre los grupos, debemos especificar qué comparaciones queremos probar.

Las comparaciones se pueden especificar utilizando la función makeContrasts.

Aquí, estamos interesados en saber qué genes se expresan diferencialmente entre los distintos grupos

Los nombres de los grupos deben coincidir exactamente con los nombres de columna de la matriz de diseño.

```
#cont.matrix <- makeContrasts(SFI-NIT,SFI-ELI,NIT-ELI,levels=design)
cont.matrix <- makeContrasts(SFIvsNIT=SFI-NIT,SFIvsELI=SFI-ELI,NITvsELI=NIT-ELI,level
s=design)
cont.matrix
```

```
##        Contrasts
## Levels SFIvsNIT SFIvsELI NITvsELI
##    ELI        0       -1       -1
##    NIT       -1        0        1
##    SFI        1        1        0
```

Las siguientes lineas se corresponden con las ya "presentadas y llevadas a cabo" y que se encuentran dentro del material de la primera parte de la asignatura.

```
fit.cont <- contrasts.fit(fit, cont.matrix)
fit.cont <- eBayes(fit.cont)
```

```
summa.fit <- decideTests(fit.cont)
summary(summa.fit)
```

```
##         SFIvsNIT SFIvsELI NITvsELI
## Down           0        0        0
## NotSig     18374    18374    18374
## Up             0        0        0
```

```
fit.cont
```

```
## An object of class "MArrayLM"
## $coefficients
##                   Contrasts
##                     SFIvsNIT   SFIvsELI   NITvsELI
##   ENSG00000227232.4 -0.1231904 -0.2343738 -0.1111834
##   ENSG00000237683.5 -0.1107212 -0.2871451 -0.1764239
##   ENSG00000241860.2 -0.1806831 -0.3124049 -0.1317218
##   ENSG00000228463.4 -0.9789662 -0.4027514  0.5762148
##   ENSG00000225972.1  1.2761546  0.9660592 -0.3100954
## 18369 more rows ...
##
## $stdev.unscaled
##                   Contrasts
##                     SFIvsNIT  SFIvsELI  NITvsELI
##   ENSG00000227232.4 0.2031788 0.1975970 0.1977131
##   ENSG00000237683.5 0.2052181 0.1983561 0.1991050
##   ENSG00000241860.2 0.4005397 0.3873543 0.3837840
##   ENSG00000228463.4 0.5154630 0.5458204 0.4811301
##   ENSG00000225972.1 0.3825753 0.3540027 0.4108064
## 18369 more rows ...
##
## $sigma
## [1] 1.008990 1.557569 0.629783 1.023444 1.817747
## 18369 more elements ...
##
## $df.residual
## [1] 27 27 27 27 27
## 18369 more elements ...
##
## $cov.coefficients
##           Contrasts
## Contrasts  SFIvsNIT SFIvsELI NITvsELI
##   SFIvsNIT      0.2      0.1     -0.1
##   SFIvsELI      0.1      0.2      0.1
##   NITvsELI     -0.1      0.1      0.2
##
## $rank
## [1] 3
##
## $Amean
## ENSG00000227232.4 ENSG00000237683.5 ENSG00000241860.2 ENSG00000228463.4
##         3.5814555         3.5493974         0.3107755        -0.7259797
## ENSG00000225972.1
##         0.4625031
## 18369 more elements ...
##
## $method
## [1] "ls"
##
## $design
##    ELI NIT SFI
## 1   0   0   1
## 2   0   0   1
## 3   0   0   1
## 4   0   0   1
## 5   0   0   1
## 25 more rows ...
```

```
##
## $contrasts
##         Contrasts
## Levels SFIvsNIT SFIvsELI NITvsELI
##    ELI        0       -1       -1
##    NIT       -1        0        1
##    SFI        1        1        0
##
## $df.prior
## [1] 3.146253
##
## $s2.prior
## [1] 0.7620268
##
## $var.prior
## [1] 0.01312290 0.01312290 0.04351578
##
## $proportion
## [1] 0.01
##
## $s2.post
## [1] 0.9913403 2.2523567 0.4347621 1.0176503 3.0388867
## 18369 more elements ...
##
## $t
##                        Contrasts
##                         SFIvsNIT   SFIvsELI   NITvsELI
##   ENSG00000227232.4 -0.6089577 -1.1912893 -0.5647978
##   ENSG00000237683.5 -0.3594982 -0.9645777 -0.5904139
##   ENSG00000241860.2 -0.6841416 -1.2231610 -0.5205292
##   ENSG00000228463.4 -1.8826560 -0.7314556  1.1871966
##   ENSG00000225972.1  1.9135029  1.5654524 -0.4330129
## 18369 more rows ...
##
## $df.total
## [1] 30.14625 30.14625 30.14625 30.14625 30.14625
## 18369 more elements ...
##
## $p.value
##                        Contrasts
##                         SFIvsNIT   SFIvsELI   NITvsELI
##   ENSG00000227232.4 0.54711446 0.2428352 0.5763903
##   ENSG00000237683.5 0.72172852 0.3424308 0.5593136
##   ENSG00000241860.2 0.49911159 0.2307447 0.6064971
##   ENSG00000228463.4 0.06942983 0.4701489 0.2444210
##   ENSG00000225972.1 0.06522173 0.1279150 0.6680878
## 18369 more rows ...
##
## $lods
##                        Contrasts
##                         SFIvsNIT   SFIvsELI   NITvsELI
##   ENSG00000227232.4 -4.687422 -4.562866 -4.883086
##   ENSG00000237683.5 -4.714943 -4.621983 -4.872106
##   ENSG00000241860.2 -4.616419 -4.577709 -4.692876
##   ENSG00000228463.4 -4.541923 -4.605211 -4.570792
##   ENSG00000225972.1 -4.498652 -4.533537 -4.690077
## 18369 more rows ...
##
```

```
## $F
## [1] 0.7029646 0.4693886 0.7450031 1.8294749 2.0990762
## 18369 more elements ...
##
## $F.p.value
## [1] 0.5030504 0.6298785 0.4832714 0.1778725 0.1401262
## 18369 more elements ...
```

```
toptable_SFIvsELI<-topTable(fit.cont,coef="SFIvsELI",sort.by="p")
toptable_SFIvsNIT<-topTable(fit.cont,coef="SFIvsNIT",sort.by="p")
toptable_NITvsELI<-topTable(fit.cont,coef="NITvsELI",sort.by="p")
```

```
# View(toptable_SFIvsELI)
# View(toptable_SFIvsNIT)
# View(toptable_NITvsELI)
```

16. Annotation and saving the results

```
library(org.Hs.eg.db)
```

```
##
```

```
columns(org.Hs.eg.db)
```

```
##  [1] "ACCNUM"       "ALIAS"        "ENSEMBL"      "ENSEMBLPROT"  "ENSEMBLTRANS"
##  [6] "ENTREZID"     "ENZYME"       "EVIDENCE"     "EVIDENCEALL"  "GENENAME"
## [11] "GO"           "GOALL"        "IPI"          "MAP"          "OMIM"
## [16] "ONTOLOGY"     "ONTOLOGYALL"  "PATH"         "PFAM"         "PMID"
## [21] "PROSITE"      "REFSEQ"       "SYMBOL"       "UCSCKG"       "UNIGENE"
## [26] "UNIPROT"
```

No he aconseguit crear les anotacions !

```
# ann <- select(org.Hs.eg.db,keys=rownames(fit.cont),columns=c("ENTREZID","SYMBOL","G
ENENAME"))
```

17. Volcano Plot

```
# We want to highlight the significant genes. We can get this from decideTests.
par(mfrow=c(1,2))
plotMD(fit.cont,coef=1,status=summa.fit[,"SFIvsNIT"], values = c(-1, 1))

# For the volcano plot we have to specify how many of the top genes to highlight.
# We can also specify that we want to plot the gene symbol for the highlighted genes.
# let's highlight the top 100 most DE genes
volcanoplot(fit.cont,coef=1,highlight=100,names=fit.cont$genes$SYMBOL)
```

SFIvsNIT
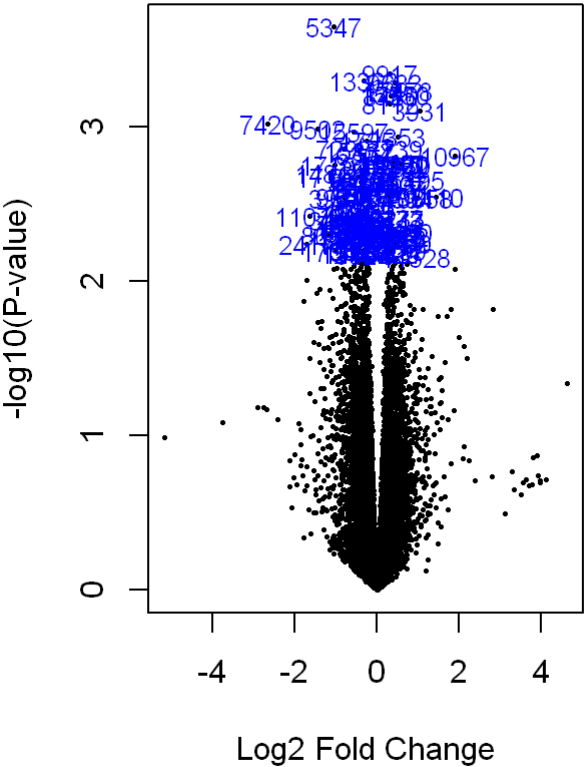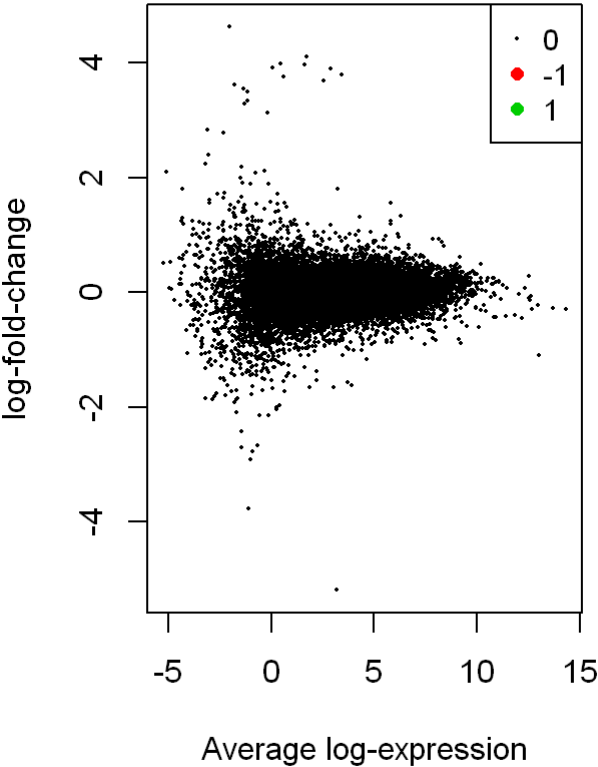
Hay una función llamada treat en el paquete limma (McCarthy y Smyth 2009) que a partir del objeto fit.conty de de un "log fold change (logFC)" determinado por el usuario como "threshold" permite "recalcular the"moderate t-statistics and p-values" . Este procedimiento es mucho más "preciso" "en el control de falsos positivos" que "listar" los p-valores y descartar a continuación genes con logFC pequeños.

```
# This is easy to do after our analysis, we just give the treat function the fit.cont
object and specify our cut-off.
fit.treat <- treat(fit.cont,lfc=1)
res.treat <- decideTests(fit.treat)
summary(res.treat)
```

```
##         SFIvsNIT SFIvsELI NITvsELI
## Down           0        0        0
## NotSig     18374    18374    18374
## Up             0        0        0
```

```
topTable(fit.treat,coef=1,sort.by="p")
```

```
##                        logFC      AveExpr          t   P.Value adj.P.Val
## ENSG00000170356.8 -2.667071 -0.6691073 -2.285105 0.01476582         1
## ENSG00000225851.1  2.828226 -3.0957945  1.661617 0.05425436         1
## ENSG00000170523.3  1.898064 -0.1426545  1.645973 0.05508328         1
## ENSG00000171195.6  4.624876 -2.0323915  1.630333 0.06515299         1
## ENSG00000160951.3  1.906968 -1.3702753  1.341059 0.09504230         1
## ENSG00000025423.7 -1.651809  3.0560536 -1.237182 0.11279568         1
## ENSG00000175535.6 -3.758024 -1.0910852 -1.319796 0.11344780         1
## ENSG00000253288.1  2.123276 -0.3166287  1.235337 0.11399672         1
## ENSG00000261600.1 -1.707952 -0.4107199 -1.230485 0.11404171         1
## ENSG00000162078.7  2.190884 -1.4227673  1.226245 0.11609204         1
```

```
# Notice that much fewer genes are highlighted in the MAplot
par(mfrow=c(1,3))
plotMD(fit.treat,coef=1,status=res.treat[,"SFIvsNIT"], values=c(-1,1))
abline(h=0,col="grey")
plotMD(fit.treat,coef=2,status=res.treat[,"SFIvsELI"], values=c(-1,1))
abline(h=0,col="grey")
plotMD(fit.treat,coef=2,status=res.treat[,"NITvsELI"], values=c(-1,1))
abline(h=0,col="grey")
```
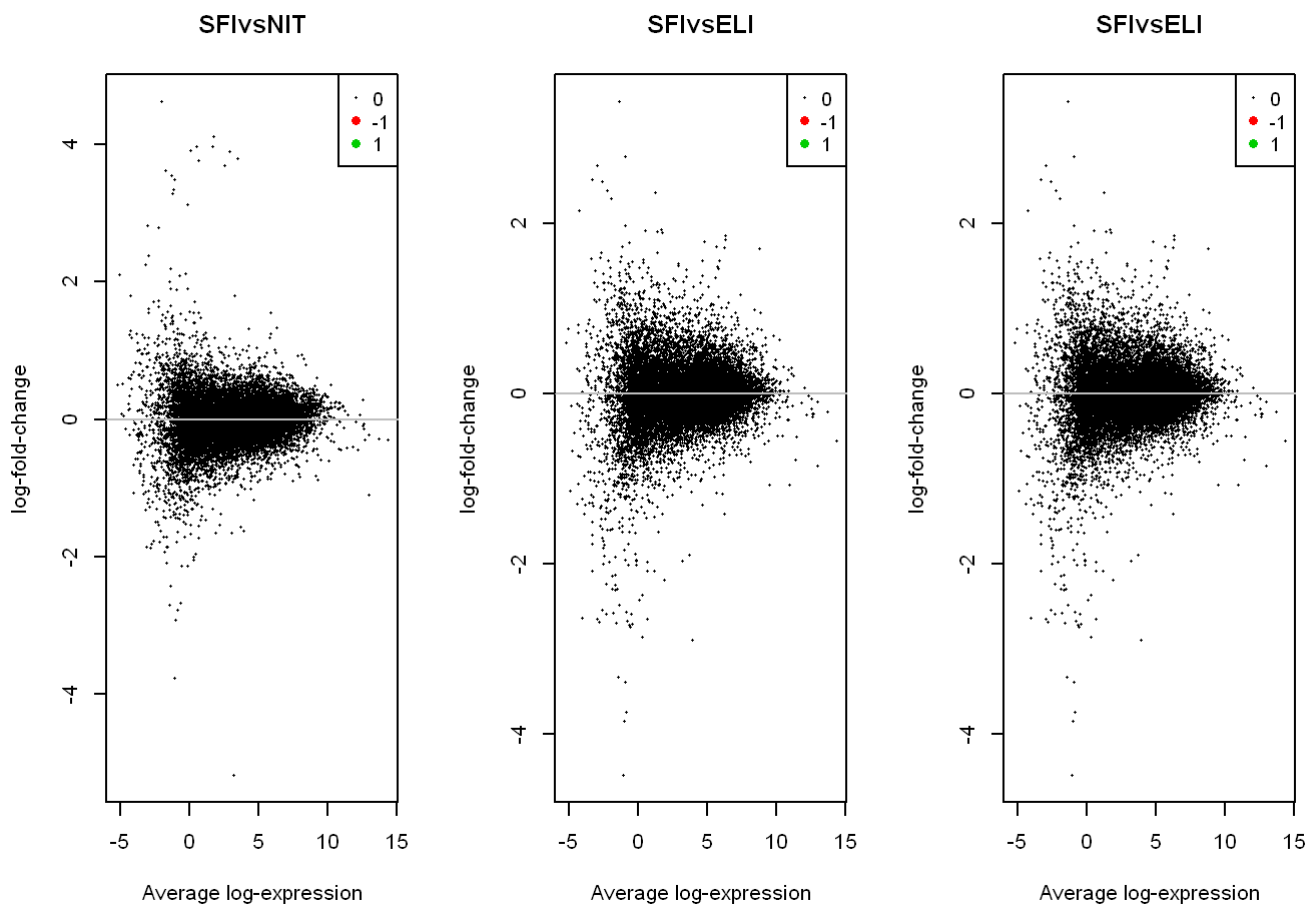

18. HeatMap

Finalmente dibujaremos el HeatMap que no se representó en el apartado 12 y quedaba pendiente.

```
## Get some nicer colours
mypalette <- brewer.pal(11,"RdYlBu")
morecols <- colorRampPalette(mypalette)
# Set up colour vector for celltype variable
col.cell <- c("purple","orange")[stargets$Group]

# Plot the heatmap
par(mfrow=c(1,1))
heatmap.2(highly_variable_lcpm,col=rev(morecols(50)),trace="none", main="Top 500 most
variable genes across samples",ColSideColors=col.cell,scale="row")
```


Top 500 most variable genes across samples
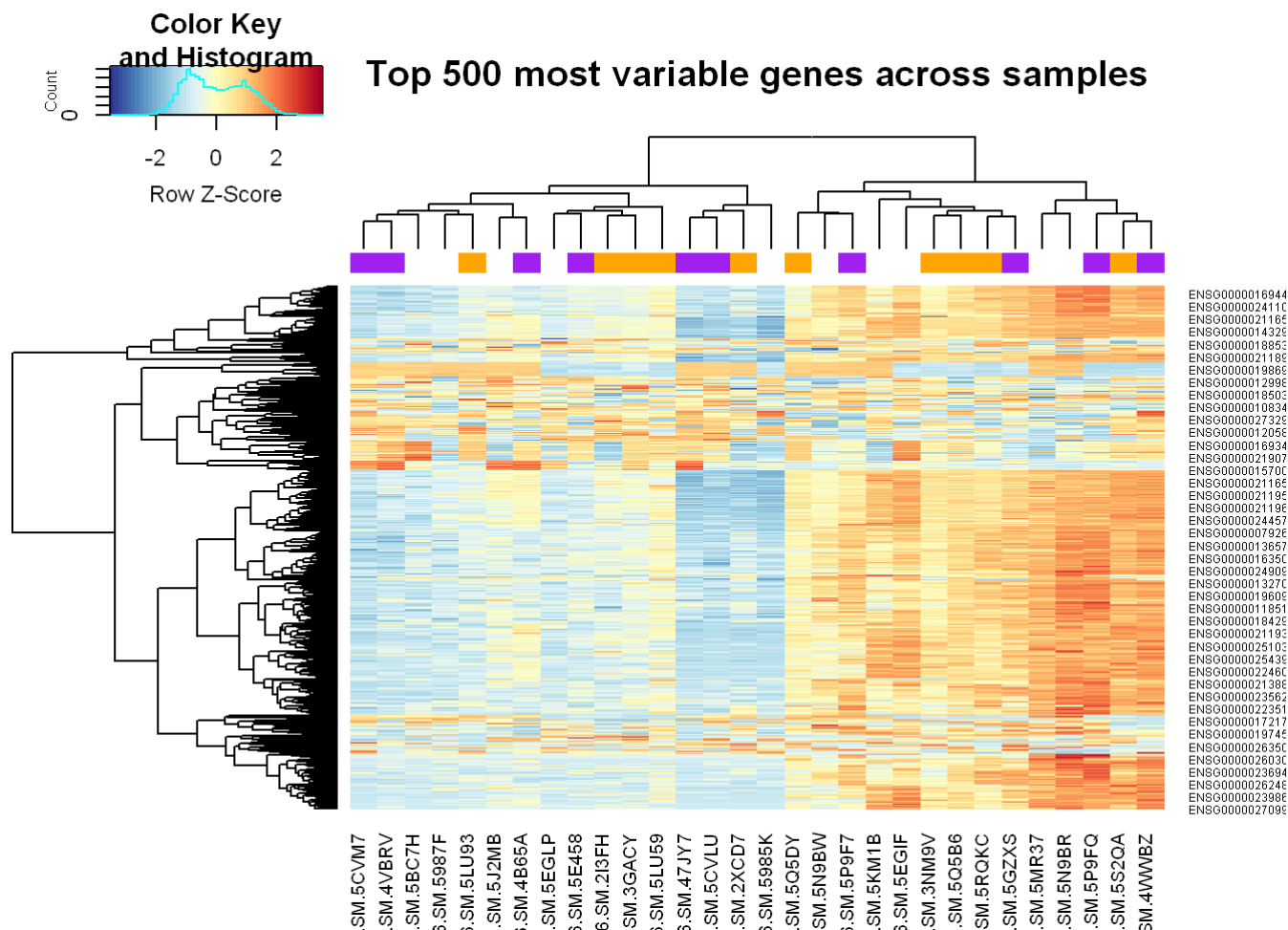
19. Referencias:

www.google.com

RNAseqTutorialUOCv2.html

Statistical analysis of RNA-seq data.pdf

IntroToAnnotationPackages.pdf

ENLACE A GitHub:

https://github.com/cmbosch/PAC2 (https://github.com/cmbosch/PAC2)