#### Full RCode

```
#LSM3241 CA1 R Code
#install Bioconductor
if (!requireNamespace("BiocManager"))
install.packages("BiocManager")
BiocManager::install()
#install Bioconductor packages
BiocManager::install('foo')
#install relevant packages needed
if (!requireNamespace("BiocManager", quietly = TRUE))
 install.packages("BiocManager")
BiocManager::install("GEOquery", version = "3.8")
BiocManager::install("affy")
BiocManager::install("limma")
BiocManager::install("hgu133plus2.db")
BiocManager::install("org.Hs.eg.db")
#load packages installed
library(affy)
library(limma)
library(hgu133plus2.db)
library(org.Hs.eg.db)
#check if packages are loaded
sessionInfo()
#Calling GSE50697
library(GEOquery)
#downloading the gse file
gse <- getGEO('GSE50697',GSEMatrix = FALSE)
#retrieving whole GSE
names(GSMList(gse))
[1] "GSM1226581" "GSM1226582" "GSM1226583" "GSM1226584" "GSM1226585"
"GSM1226586"
```

#### **#Getting the raw data for the series using GEOquery** filePaths <- getGEOSuppFiles('GSE50697') #generating a vector containing the name of all the CEL files. list.celfiles('GSE50697\_RAW')

- [1] "GSM1226581\_S1100958.MDA.01\_HG-U133\_Plus\_2\_.CEL.gz"
- [2] "GSM1226582\_S1100958.MDA.02\_HG-U133\_Plus\_2\_.CEL.gz"
- [3] "GSM1226583\_S1100958.MDA.03\_HG-U133\_Plus\_2\_.CEL.gz"
- [4] "GSM1226584\_S1100958.MDA.04\_HG-U133\_Plus\_2\_.CEL.gz"
- [5] "GSM1226585\_S1100958.MDA.05\_HG-U133\_Plus\_2\_.CEL.gz"
- [6] "GSM1226586\_S1100958.MDA.06\_HG-U133\_Plus\_2\_.CEL.gz"

## #finding info from meta data of gse

# names(Meta(gse))

[41    4 4   4	11 4 4 14 - 11	U = 4 = - 4 =
[1] "contact_address"	"contact_city"	"contact_country"
[4] "contact_department"	"contact_email"	"contact_institute'
[7] "contact_laboratory"	"contact_name"	"contact_phone"
[10] "contact_state"	"contact_zip/postal_code"	"contributor"
[13] "email"	"geo_accession"	"institute"
[16] "last_update_date"	"name"	"overall_design"
[19] "platform_id"	"platform_taxid"	"pubmed_id"
[22] "relation"	"sample_id"	"sample_taxid"

[22] "relation" 'sample\_id" [25] "status" "submission\_date" "summary" "title" "type"

[28] "supplementary\_file"

[31] "web\_link"

## #creating GSM object names

gsm <- GSMList(gse)[[1]]

#### #find information we need from the metadata of gsm names(Meta(gsm))

[1] "channel_count"	"characteristics_ch1"	"contact_address"
[4] "contact_city"	"contact_country"	"contact_department"
[7] "contact_email"	"contact_institute"	"contact_laboratory"
[10] "contact_name"	"contact_phone"	"contact_state"
[13] "contact_zip/postal_code"	"data_processing"	"data_row_count"
[16] "extract_protocol_ch1"	"geo_accession"	"hyb_protocol"
[19] "label_ch1"	"label_protocol_ch1"	"last_update_date"
[22] "molecule_ch1"	"organism_ch1"	"platform_id"

[25] "scan\_protocol""series\_id""source\_name\_ch1"[28] "status""submission\_date""supplementary\_file"

[31] "taxid\_ch1" "title" "type"

#### #extract metadata of gsm not in gse

#### names(Meta(gsm))[!(names(Meta(gsm)) %in% names(Meta(gse)))]

[1] "channel\_count""characteristics\_ch1" "data\_processing"[4] "data\_row\_count""extract\_protocol\_ch1""hyb\_protocol"[7] "label\_ch1""label\_protocol\_ch1""molecule\_ch1"[10] "organism\_ch1""scan\_protocol""series\_id"

[13] "source\_name\_ch1" "taxid\_ch1"

#### Meta(gsm)[!(names(Meta(gsm)) %in% names(Meta(gse)))]

\$channel\_count

[1] "1"

\$characteristics\_ch1

[1] "cell line: SUM159" "treatment: control"

[3] "tissue: Claudin-low breast cancer"

\$data\_processing

[1] "MAS 5.0"

\$data\_row\_count

[1] "54675"

\$extract\_protocol\_ch1

[1] "standard Affymetrix protocol"

\$hyb\_protocol

[1] "standard Affymetrix protocol"

\$label\_ch1

[1] "biotin"

\$label\_protocol\_ch1

[1] "standard Affymetrix protocol"

\$molecule\_ch1

[1] "total RNA"

```
$organism ch1
[1] "Homo sapiens"
$scan protocol
[1] "standard Affymetrix protocol"
$series id
[1] "GSE50697"
$source_name_ch1
[1] "SUM159"
$taxid_ch1
[1] "9606"
#From the metadata information, we decided to use the following elements
#source_name_ch1 and characteristics_ch1
#getting sample growth conditions into a data frame
culture_medium <- function(gsm) {</pre>
 Meta(gsm)[['characteristics_ch1']][2]
}
sapply(GSMList(gse),culture_medium)
             GSM1226581
                                       GSM1226582
          "treatment: control"
                                  "treatment: control"
             GSM1226583
                                       GSM1226584
      "treatment: control"
                                  "treatment: pBabe puro miR-203"
             GSM1226585
                                       GSM1226586
"treatment: pBabe puro miR-203"
                                  "treatment: pBabe puro miR-203"
pd <- data.frame(culture=as.factor(sapply(GSMList(gse),culture_medium)))</pre>
pd
                           culture
GSM1226581
                       treatment: control
GSM1226582
                       treatment: control
GSM1226583
                       treatment: control
GSM1226584 treatment: pBabe puro miR-203
GSM1226585 treatment: pBabe puro miR-203
GSM1226586 treatment: pBabe puro miR-203
#simplifying our dataframe, convert columns to 2 values only
pd$culture <- as.factor(pd$culture)
levels(pd$culture) <- c("control","miR203")</pre>
```

#to enable the kable function, we need to install the knitr package install.packages("knitr")
library(knitr)
kable(pd)

```
| | |culture |
|:-----|
|GSM1226581 |control |
|GSM1226582 |control |
|GSM1226583 |control |
|GSM1226584 |miR203 |
|GSM1226585 |miR203 |
|GSM1226586 |miR203 |
```

#### #Reading in the CEL files with the phenoData

celfiles <- paste0('GSE50697\_RAW/', list.celfiles('GSE50697\_RAW/'),'.')
affydata <- read.affybatch(celfiles,phenoData = new("AnnotatedDataFrame",pd))
phenoData(affydata)

An object of class 'AnnotatedDataFrame'

sampleNames: GSM1226581 GSM1226582 ... GSM1226586 (6 total)

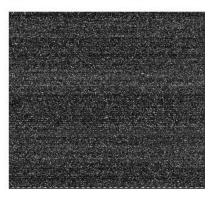
varLabels: culture

varMetadata: labelDescription

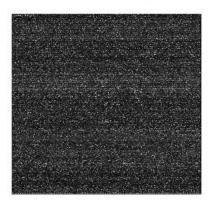
#### #pseudo images of chips

image(data[,1])

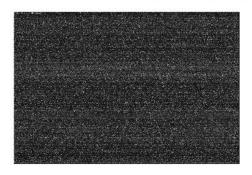
1226581 S1100958.MDA.01 HG-U133 Plus



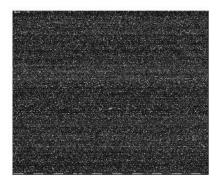
image(data[,2]) 26582\_S1100958.MDA.02\_HG-U133\_PI



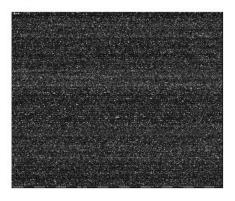
image(data[,3])
GSM1226583\_S1100958.MDA.03\_HG-U133\_Plus\_2\_.(



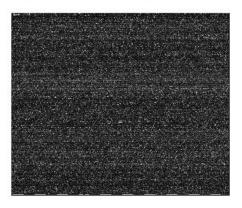
image(data[,4]) W1226584\_S1100958.MDA.04\_HG-U133\_Plus\_



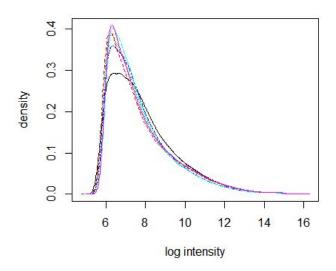
image(data[,5]) W1226585\_S1100958.MDA.05\_HG-U133\_Plus\_



image(data[,6]) W1226586\_S1100958.MDA.06\_HG-U133\_Plus\_

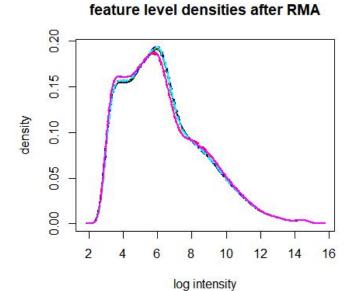


**#CEL file densities before normalisation or background correction plotDensity.AffyBatch(affydata)** 



#### **#perform RMA normalisation**

eset <- rma(affydata)
plotDensity(exprs(eset),xlab='log intensity',main="feature level densities after
RMA",lwd=2)</pre>



# #phenotype data for each samples after rma is retained pData(eset)

culture GSM1226581 control GSM1226582 control GSM1226583 control GSM1226584 miR203 GSM1226585 miR203 GSM1226586 miR203

#### #generate model matrix

model <- model.matrix( ~ 0 + eset\$culture)
#rename the model columns to correspond to the different growth conditions
colnames(model) <- levels(eset\$culture)
model

```
control miR203
1
    1
         0
         0
2
    1
    1 0
3
4
    0 1
     0 1
5
     0 1
attr(,"assign")
[1] 1 1
attr(,"contrasts")
attr(,"contrasts")$`eset$culture`
[1] "contr.treatment"
```

#look at when the growth conditions differ, we create contrast contrasts <- makeContrasts(control - miR203, levels=model) contrasts

Contrasts

Levels control - miR203

control 1

miR203 -1

#fit model and contrast matrix into a data

fit <- ImFit(eset, model)

#### fit

#### An object of class "MArrayLM" \$coefficients

	control	miR203		
1007_s_at	9.245338	9.408269		
1053_at	9.586208	9.605038		
117_at	6.404494	6.500503		
121_at	8.496605	8.577745		
1255 <u>g</u> at	3.425654	3.360337		
54670 more rows				

#### \$rank

[1] 2

#### \$assign

[1] 1 1

#### \$qr

\$qr

control miR203 1 -1.7320508 0.0000000 2 0.5773503 -1.7320508 3 0.5773503 0.0000000 4 0.0000000 0.5773503 5 0.0000000 0.5773503 6 0.0000000 0.5773503 attr(,"assign")

[1] 1 1

attr(,"contrasts")

attr(,"contrasts")\$`eset\$culture`

[1] "contr.treatment"

#### \$qraux

[1] 1.57735 1.00000

#### \$pivot

[1] 1 2

#### \$tol

[1] 1e-07

#### \$rank

[1] 2

#### \$df.residual

[1] 4 4 4 4 4

54670 more elements ...

#### \$sigma

1007\_s\_at 1053\_at 117\_at 121\_at 1255\_g\_at 0.07378465 0.13302616 0.18946888 0.14112115 0.06639880 54670 more elements ...

#### \$cov.coefficients

control miR203 control 0.3333333 0.0000000 miR203 0.0000000 0.3333333

#### \$stdev.unscaled

	control	miR203		
1007_s_at	0.5773503	0.5773503		
1053_at	0.5773503	0.5773503		
117_at	0.5773503	0.5773503		
121_at	0.5773503	0.5773503		
1255_g_at	0.5773503	0.5773503		
54670 more rows				

0.0.0.0.0.0.0.0.0.0.0

#### \$pivot

[1] 1 2

#### \$Amean

1007\_s\_at 1053\_at 117\_at 121\_at 1255\_g\_at 9.326804 9.595623 6.452499 8.537175 3.392996 54670 more elements ...

#### \$method

[1] "Is"

#### \$design

control miR203

- 1 1 0
- 2 1 0
- 3 1 0

```
0 1
4
5
     0 1
6
         1
attr(,"assign")
[1] 1 1
attr(,"contrasts")
attr(,"contrasts")$`eset$culture`
[1] "contr.treatment"
fitted.contrast <- contrasts.fit(fit,contrasts)</pre>
fitted.contrast
An object of class "MArrayLM"
$coefficients
      Contrasts
              control - miR203
 1007_s_at -0.16293107
 1053_at
              -0.01883039
 117_at
              -0.09600894
 121_at
              -0.08113934
 1255_g_at
               0.06531688
54670 more rows ...
$rank
[1] 2
$assign
[1] 1 1
$qr
$qr
   control
              miR203
1 -1.7320508 0.0000000
2 0.5773503 -1.7320508
3 0.5773503 0.0000000
4 0.0000000 0.5773503
5 0.0000000 0.5773503
6 0.0000000 0.5773503
attr(,"assign")
[1] 1 1
attr(,"contrasts")
attr(,"contrasts")$`eset$culture`
[1] "contr.treatment"
```

```
$qraux
```

[1] 1.57735 1.00000

\$pivot

[1] 1 2

\$tol

[1] 1e-07

\$rank

[1] 2

#### \$df.residual

[1] 4 4 4 4 4

54670 more elements ...

#### \$sigma

1007\_s\_at 1053\_at 117\_at 121\_at 1255\_g\_at 0.07378465 0.13302616 0.18946888 0.14112115 0.06639880 54670 more elements ...

#### \$cov.coefficients

Contrasts

Contrasts control - miR203 control - miR203 0.6666667

#### \$stdev.unscaled

#### Contrasts

control - miR203

 1007\_s\_at
 0.8164966

 1053\_at
 0.8164966

 117\_at
 0.8164966

 121\_at
 0.8164966

 1255\_g\_at
 0.8164966

54670 more rows ...

#### \$Amean

1007\_s\_at 1053\_at 117\_at 121\_at 1255\_g\_at 9.326804 9.595623 6.452499 8.537175 3.392996 54670 more elements ...

#### \$method

[1] "ls"

#### \$design

#### \$contrasts

Contrasts

Levels control - miR203 control 1 miR203 -1

# #calculate test statistics via performing eBayes correction fitted.ebayes <- eBayes(fitted.contrast) fitted.ebayes

An object of class "MArrayLM" \$coefficients

#### Contrasts

control - miR203 1007\_s\_at -0.16293107 1053\_at -0.01883039 117\_at -0.09600894 121\_at -0.08113934 1255\_g\_at 0.06531688 54670 more rows ...

#### \$rank

[1] 2

```
[1] 1 1
$qr
$qr
   control
             miR203
1 -1.7320508 0.0000000
2 0.5773503 -1.7320508
3 0.5773503 0.0000000
4 0.0000000 0.5773503
5 0.0000000 0.5773503
6 0.0000000 0.5773503
attr(,"assign")
[1] 1 1
attr(,"contrasts")
attr(,"contrasts")$`eset$culture`
[1] "contr.treatment"
$qraux
[1] 1.57735 1.00000
$pivot
[1] 1 2
$tol
[1] 1e-07
$rank
[1] 2
$df.residual
[1] 4 4 4 4 4
54670 more elements ...
$sigma
1007_s_at 1053_at 117_at 121_at 1255_g_at
0.07378465\ 0.13302616\ 0.18946888\ 0.14112115\ 0.06639880
54670 more elements ...
$cov.coefficients
```

\$assign

```
Contrasts
```

Contrasts control - miR203 control - miR203 0.6666667

#### \$stdev.unscaled

#### Contrasts

control - miR203

 1007\_s\_at
 0.8164966

 1053\_at
 0.8164966

 117\_at
 0.8164966

 121\_at
 0.8164966

 1255\_g\_at
 0.8164966

54670 more rows ...

#### \$Amean

1007\_s\_at 1053\_at 117\_at 121\_at 1255\_g\_at 9.326804 9.595623 6.452499 8.537175 3.392996 54670 more elements ...

#### \$method

[1] "ls"

#### \$design

#### control miR203

1 1 0

2 1 0

3 1 0

4 0 1

5 0 1

6 0 1

attr(,"assign")

[1] 1 1

attr(,"contrasts")

attr(,"contrasts")\$`eset\$culture`

[1] "contr.treatment"

#### \$contrasts

#### Contrasts

Levels control - miR203

control 1 miR203 -1

```
$df.prior
[1] 5.163163
$s2.prior
[1] 0.01315895
$var.prior
[1] 51.55685
$proportion
[1] 0.01
$s2.post
 1007_s_at 1053_at 117_at 121_at 1255_g_at
0.009791214 0.015139491 0.023085437 0.016108249 0.009339242
54670 more elements ...
$t
      Contrasts
             control - miR203
 1007_s_at -2.0166534
 1053_at -0.1874344
117_at -0.7739061
 121_at
             -0.7829841
 1255_g_at
              0.8277806
54670 more rows ...
$df.total
[1] 9.163163 9.163163 9.163163 9.163163 9.163163
54670 more elements ...
$p.value
      Contrasts
             control - miR203
 1007_s_at 0.07396289
 1053_at
            0.85540486
           0.45848725
 117_at
 121_at
             0.45339651
 1255 g at
             0.42882619
54670 more rows ...
```

\$lods

Contrasts

```
control - miR203
```

 1007\_s\_at
 -4.937866

 1053\_at
 -6.756422

 117\_at
 -6.458114

 121\_at
 -6.450863

 1255\_g\_at
 -6.414005

54670 more rows ...

#### \$F

[1] 4.06689075 0.03513165 0.59893067 0.61306414 0.68522080 54670 more elements ...

#### \$F.p.value

[1] 0.07396289 0.85540486 0.45848725 0.45339651 0.42882619 54670 more elements ...

#### #extracting differentially expressed genes

#### topTable(fitted.ebayes)

logFC AveExpr t P.Value 209719\_x\_at -2.096174 9.668246 -27.98136 3.449900e-10 210413 x at -1.984099 9.649669 -25.77291 7.265063e-10 201721 s at 2.254152 9.710288 25.72106 7.398668e-10 211906\_s\_at -2.143763 9.071134 -24.70101 1.066850e-09 206172\_at -1.789371 7.706033 -24.07788 1.343931e-09 209720 s at -2.432950 8.729486 -23.12911 1.931967e-09 211756 at -3.067195 4.671276 -21.91533 3.140852e-09 202949\_s\_at -1.502988 10.092484 -20.84188 4.935440e-09 202007\_at 3.494914 5.646599 20.62198 5.429179e-09 206002 at 1.704752 8.907492 20.45564 5.839018e-09 adj.P.Val В 209719\_x\_at 1.348407e-05 12.13796 210413 x at 1.348407e-05 11.73165 201721 s at 1.348407e-05 11.72129 211906\_s\_at 1.458251e-05 11.50898 206172 at 1.469588e-05 11.37091 209720 s at 1.760505e-05 11.14745 211756 at 2.453230e-05 10.83610 202949\_s\_at 3.192483e-05 10.53422 202007 at 3.192483e-05 10.46905 206002 at 3.192483e-05 10.41895

#### #Get a limited number of probesets

## ps <- rownames(topTable(fitted.ebayes)) ps</pre>

- [1] "209719\_x\_at" "210413\_x\_at" "201721\_s\_at" "211906\_s\_at"
- [5] "206172 at" "209720 s at" "211756 at" "202949 s at"
- [9] "202007\_at" "206002\_at"

#### **#The AnnotationDbi interface**

#look at available columns for our chip

#### columns(hgu133plus2.db)

- [1] "ACCNUM" "ALIAS" "ENSEMBL" "ENSEMBLPROT"
- [5] "ENSEMBLTRANS" "ENTREZID" "ENZYME" "EVIDENCE"
- [9] "EVIDENCEALL" "GENENAME" "GO" "GOALL"
- [13] "IPI" "MAP" "OMIM" "ONTOLOGY"
- [17] "ONTOLOGYALL" "PATH" "PFAM" "PMID"
- [21] "PROBEID" "PROSITE" "REFSEQ" "SYMBOL"
- [25] "UCSCKG" "UNIGENE" "UNIPROT"

#### #Look at which that can be used as keys

#### keytypes(hgu133plus2.db)

- [1] "ACCNUM" "ALIAS" "ENSEMBL" "ENSEMBLPROT"
- [5] "ENSEMBLTRANS" "ENTREZID" "ENZYME" "EVIDENCE"
- [9] "EVIDENCEALL" "GENENAME" "GO" "GOALL"
- [13] "IPI" "MAP" "OMIM" "ONTOLOGY"
- [17] "ONTOLOGYALL" "PATH" "PFAM" "PMID"
- [21] "PROBEID" "PROSITE" "REFSEQ" "SYMBOL"
- [25] "UCSCKG" "UNIGENE" "UNIPROT"

## #realise all can be used as keys, choose "PROBEID" as most suitable head(keys(hgu133plus2.db,keytype="PROBEID"))

```
[1] "1007_s_at" "1053_at" "117_at" "121_at" "1255_g_at" [6] "1294_at"
```

## AnnotationDbi::select(hgu133plus2.db,ps,c("SYMBOL","ENTREZID","GENENAME"),keyty pe="PROBEID")

PROBEID	SYMBOL	ENTREZID
1 209719_x_at	SERPINB3	6317
2 210413_x_at	SERPINB4	6318
3 210413 x at	SERPINB3	6317

```
4 201721 s at
                   LAPTM5
                                7805
5 211906_s_at
                   SERPINB4
                                6318
6 206172 at
                   IL13RA2
                                3598
7 209720_s_at
                   SERPINB3
                                6317
8 211756 at
                   PTHLH
                                5744
9 202949_s_at
                   FHL2
                                2274
10 202007 at
                   NID1
                                4811
11 206002 at
                   ADGRG2
                                10149
                   GENENAME
1
          serpin family B member 3
2
          serpin family B member 4
3
          serpin family B member 3
4
     lysosomal protein transmembrane 5
5
          serpin family B member 4
6 interleukin 13 receptor subunit alpha 2
7
          serpin family B member 3
8
      parathyroid hormone like hormone
9
       four and a half LIM domains 2
10
                   nidogen 1
11 adhesion G protein-coupled receptor G2
```

#### #Restrict to upregulated genes

differentially\_expressed <- topTable(fitted.ebayes,number=Inf,p.value=0.1,Ifc=1) differentially\_expressed

```
loaFC
                      AveExpr
                                   t
                                           P.Value
209719_x_at -2.096174 9.668246 -27.981362 3.449900e-10
210413 x at -1.984099 9.649669 -25.772906 7.265063e-10
201721 s at 2.254152 9.710288 25.721059 7.398668e-10
211906 s at -2.143763 9.071134 -24.701007 1.066850e-09
206172 at -1.789371 7.706033 -24.077878 1.343931e-09
209720 s at -2.432950 8.729486 -23.129114 1.931967e-09
211756 at -3.067195 4.671276 -21.915333 3.140852e-09
202949 s at -1.502988 10.092484 -20.841881 4.935440e-09
202007 at 3.494914 5.646599 20.621979 5.429179e-09
206002 at 1.704752 8.907492 20.455638 5.839018e-09
228946 at 2.850568 7.433498 19.353438 9.598769e-09
203296_s_at 1.436208 7.128925 18.624289 1.353915e-08
200953 s at 2.114548 6.413324 18.518165 1.424938e-08
201681 s at 1.343568 10.848433 18.048727 1.792645e-08
200618 at 1.396831 10.601469 17.304137 2.610933e-08
214051 at 1.266167 6.486634 16.511752 3.963213e-08
```

```
205207 at -1.972971 10.543937 -16.302947 4.437863e-08
215785 s at 1.310530 6.548376 16.145121 4.838383e-08
218499 at 2.660353 6.810620 16.064949 5.057022e-08
225295 at 1.134359 9.370925 15.777075 5.936971e-08
210445 at 1.244751 7.713232 15.766920 5.970961e-08
225237_s_at 1.424176 7.916869 15.762484 5.985878e-08
204602_at -1.266821 11.377001 -15.706611 6.177312e-08
204019 s at 1.947556 8.145434 15.701004 6.196895e-08
201739 at -1.138479 9.582154 -15.278721 7.889207e-08
222288 at 1.827582 7.897300 14.987890 9.350461e-08
217388 s at -1.340457 9.048417 -14.945053 9.589930e-08
1556773 at -1.928811 5.156095 -14.816192 1.035213e-07
227609_at -1.245758 7.672937 -14.484025 1.264433e-07
215189 at 1.468563 10.044900 14.282840 1.430238e-07
222857 s at -1.132656 8.121989 -14.208997 1.497009e-07
213711 at 1.159206 9.053300 14.131621 1.570690e-07
241864 x at 1.489112 7.380173 14.128284 1.573956e-07
209946 at -1.781273 6.908853 -13.616536 2.176355e-07
226134 s at 1.136151 9.222280 13.608353 2.187858e-07
219049 at -1.948935 5.572982 -13.575335 2.234962e-07
212230 at 1.313188 8.909360 13.492343 2.358382e-07
202242 at 1.407202 5.471476 13.412674 2.483975e-07
206343 s at -1.609212 8.373871 -13.352914 2.583030e-07
209739 s at 1.121705 8.529516 13.309525 2.657682e-07
218162 at 1.045352 6.832097 13.283405 2.703769e-07
207802 at 1.667464 5.074460 13.166901 2.920360e-07
1557883 a at 1.806322 4.515782 12.969677 3.332030e-07
224901 at 1.490742 6.113577 12.888866 3.518878e-07
201720 s at 1.174496 8.359157 12.888647 3.519399e-07
212543 at -1.530326 6.498998 -12.838209 3.641887e-07
227963 at 1.444880 6.061685 12.834804 3.650323e-07
226571 s at 1.196636 8.630486 12.813410 3.703829e-07
213436 at 1.271657 9.635054 12.756578 3.850216e-07
1553995 a at -1.253015 7.654807 -12.658433 4.118291e-07
203910 at -1.315174 9.064571 -12.634820 4.185819e-07
227156 at 1.328361 5.991962 12.582181 4.340796e-07
228863 at 1.472731 6.357374 12.506484 4.574876e-07
223395 at 1.068056 10.465127 12.441449 4.787137e-07
206068_s_at 1.424994 5.780641 12.400733 4.925544e-07
208891 at -1.385639 8.398753 -12.318390 5.219150e-07
212942 s at -1.516025 6.738569 -12.259489 5.441012e-07
1553602 at 1.363603 6.596712 12.236775 5.529329e-07
205534_at -1.336143 8.974132 -12.149474 5.883802e-07
```

```
205767 at -1.716322 8.767721 -12.118055 6.017450e-07
228742 at 1.041995 7.583651 12.046972 6.332343e-07
223557 s at -1.363652 7.888034 -12.009754 6.504438e-07
235274 at 1.005532 7.657250 11.748894 7.865815e-07
216870 x at -1.296386 6.322829 -11.714764 8.066076e-07
211964_at -1.063786 6.932677 -11.585661 8.876065e-07
225240 s at 1.297440 8.599465 11.531470 9.242444e-07
243309 at 1.127696 5.892718 11.521002 9.315123e-07
213947 s at 1.096488 7.688141 11.481711 9.593578e-07
219874 at -1.032381 7.306612 -11.477603 9.623221e-07
243880 at 1.122249 5.598890 11.461119 9.743176e-07
227020 at -1.008496 8.595696 -11.460947 9.744438e-07
224856 at 1.191607 9.560194 11.401104 1.019403e-06
207623 at 1.160545 7.651684 11.384154 1.032549e-06
205404 at -1.277894 6.526299 -11.351684 1.058258e-06
202712 s at 1.239059 7.502293 11.345660 1.063105e-06
205110 s at -1.128768 9.124357 -11.343864 1.064554e-06
226463 at 1.665975 7.020946 11.258157 1.136338e-06
1560225 at 1.186996 6.322784 11.230714 1.160437e-06
217996 at -1.340874 9.870791 -11.174406 1.211669e-06
217997 at -1.471840 8.227516 -11.142785 1.241528e-06
221731 x at -1.255271 7.842270 -11.133802 1.250158e-06
214803 at 1.349396 6.911944 11.108018 1.275297e-06
213201_s_at 1.095576 7.502540 11.021185 1.364143e-06
210663 s at -1.162039 8.866615 -11.019233 1.366217e-06
226921 at 1.040167 8.608740 11.009422 1.376696e-06
224691 at 1.015055 10.545175 11.003052 1.383545e-06
222920_s_at 1.051959 6.425689 10.994649 1.392639e-06
208892 s at -1.349587 7.710335 -10.978803 1.409967e-06
227061 at 1.270223 7.406550 10.851655 1.557932e-06
201150 s at 1.744768 7.016096 10.795555 1.628587e-06
203836 s at -1.173029 6.334483 -10.701144 1.755598e-06
226757 at -1.156531 7.348931 -10.666205 1.805336e-06
205476 at -1.026426 9.118523 -10.493921 2.074396e-06
209355_s_at 1.283780 7.946635 10.412755 2.216216e-06
203837 at -1.664224 6.336915 -10.408882 2.223246e-06
204194 at -1.476294 7.693366 -10.378921 2.278458e-06
213316_at 1.024644 5.038984 10.317175 2.397069e-06
226189_at -1.494698 6.607014 -10.265638 2.501282e-06
215629 s at -1.390720 5.298695 -10.262250 2.508306e-06
210503 at 1.028117 7.913849 10.241691 2.551394e-06
227486 at -1.368408 7.642176 -10.239413 2.556218e-06
214467_at -1.382421 4.825823 -10.213686 2.611405e-06
```

```
235199 at -1.550341 6.769346 -10.167323 2.714197e-06
209687 at -1.177895 6.825387 -10.165181 2.719053e-06
205003 at -1.531919 6.640436 -10.141147 2.774194e-06
227289 at 1.331888 5.596073 10.084716 2.908564e-06
204620 s at -1.035997 7.368166 -10.031903 3.040828e-06
226370 at 1.157409 7.227893 10.010942 3.095146e-06
210118 s at -1.458621 7.811831 -9.897507 3.408250e-06
204385 at -1.010836 8.197030 -9.896692 3.410621e-06
204533 at -1.103127 6.696345 -9.846135 3.561377e-06
226302 at -1.190951 9.588665 -9.774950 3.786216e-06
213391 at 1.199648 7.855011 9.693048 4.064410e-06
206290 s at -1.210959 6.322545 -9.671217 4.142303e-06
204114_at -1.255358 4.847498 -9.647616 4.228366e-06
211506 s at -1.199992 10.684143 -9.621588 4.325565e-06
228033 at -1.075819 8.476069 -9.596823 4.420332e-06
223869 at -1.461387 9.000998 -9.519697 4.730392e-06
224938 at 1.003431 8.507051 9.506805 4.784520e-06
238937 at 1.512821 5.334978 9.481948 4.890806e-06
219287 at -1.150889 7.041430 -9.444304 5.056746e-06
226218 at -1.291518 7.704088 -9.321460 5.642803e-06
239866 at 1.021869 7.209041 9.298515 5.760337e-06
240391 at 1.095573 5.639588 9.267364 5.924232e-06
227875 at 1.088525 5.477780 9.240871 6.067649e-06
212226 s at 1.141638 8.813665 9.174575 6.443553e-06
1557882 at 1.308155 4.398610 8.996530 7.585896e-06
1553994 at -1.195990 8.464710 -8.929526 8.071966e-06
222446 s at 1.109060 6.870658 8.787498 9.219349e-06
238953 at 1.102176 7.213338 8.767421 9.395512e-06
227480 at -1.008557 6.454705 -8.737691 9.663203e-06
211122 s at -1.437891 5.481624 -8.701734 9.998195e-06
204597 x at -1.114211 6.475372 -8.669453 1.030983e-05
1552546 a at -1.056041 7.462691 -8.640677 1.059662e-05
226492 at 1.047015 6.477821 8.578079 1.125120e-05
208893 s at -1.003504 7.551996 -8.572491 1.131176e-05
1553684_at 1.159679 6.705608 8.363742 1.385147e-05
204619 s at -1.318002 6.670587 -8.300155 1.474473e-05
238462 at -1.140790 6.530780 -8.119298 1.764965e-05
219961 s at 1.026839 9.636927 8.032967 1.925292e-05
231568_at -1.111633 5.995260 -7.998804 1.993085e-05
1557094 at -1.013669 8.179558 -7.922769 2.153554e-05
210941 at -1.137155 4.912168 -7.864326 2.286515e-05
228450_at -1.335993 5.632521 -7.829498 2.370007e-05
209839_at -1.148218 6.531914 -7.736205 2.610565e-05
```

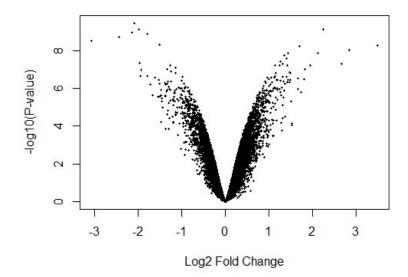
207173 x at -1.212647 6.236454 -7.597516 3.019017e-05 211959\_at 1.364346 6.466787 7.528684 3.247290e-05 222662 at -1.274162 6.655163 -7.401749 3.719376e-05 1569157 s at 1.318162 5.058315 7.357447 3.901403e-05 224367 at 1.063579 6.028615 7.175741 4.756705e-05 205830\_at 1.179054 5.879032 7.154081 4.871626e-05 212706 at 1.105566 4.929178 7.120332 5.056777e-05 205535 s at -1.191363 7.219878 -6.967438 5.997430e-05 210355 at -1.479006 4.670382 -6.962900 6.028118e-05 205532 s at 1.274534 6.604113 6.910359 6.396223e-05 1558501 at -1.198424 5.294933 -6.885537 6.578596e-05 201147 s at 1.530476 6.170983 6.812613 7.148008e-05 1555673 at -1.051042 6.785199 -6.792971 7.310412e-05 235561 at 1.079808 5.430246 6.769854 7.506718e-05 202920 at 1.527176 5.747694 6.668522 8.437675e-05 243009 at 1.039904 4.651447 6.630249 8.821433e-05 220030 at -1.294684 4.335316 -6.250504 1.384963e-04 206300 s at -1.374347 4.566611 -6.212739 1.449942e-04 205533 s at 1.035088 5.136024 6.189540 1.491497e-04 220301 at 1.291894 5.727456 6.061185 1.746081e-04 1554131 at -1.084905 4.428248 -5.667805 2.868936e-04 adj.P.Val В 209719\_x\_at 1.348407e-05 12.1379607 210413\_x\_at 1.348407e-05 11.7316498 201721\_s\_at 1.348407e-05 11.7212891 211906 s at 1.458251e-05 11.5089762 206172 at 1.469588e-05 11.3709081 209720\_s\_at 1.760505e-05 11.1474545 211756 at 2.453230e-05 10.8360961 202949 s at 3.192483e-05 10.5342229 202007 at 3.192483e-05 10.4690463 206002 at 3.192483e-05 10.4189530 228946 at 4.771024e-05 10.0688951 203296 s at 5.992960e-05 9.8187527 200953\_s\_at 5.992960e-05 9.7810275 201681 s at 7.000917e-05 9.6099414 200618 at 9.516851e-05 9.3238050 214051 at 1.354304e-04 8.9978822 205207 at 1.411730e-04 8.9080757 215785 s at 1.411730e-04 8.8390616 218499 at 1.411730e-04 8.8036239 225295 at 1.411730e-04 8.6742241 210445\_at 1.411730e-04 8.6695970

225237 s at 1.411730e-04 8.6675742 204602\_at 1.411730e-04 8.6420285 204019 s at 1.411730e-04 8.6394576 201739 at 1.725370e-04 8.4419789 222288 at 1.941961e-04 8.3014157 217388 s at 1.941961e-04 8.2803884 1556773\_at 2.021437e-04 8.2166245 227609 at 2.383893e-04 8.0486591 215189 at 2.606610e-04 7.9443366 222857 s at 2.607759e-04 7.9055435 213711 at 2.607759e-04 7.8646006 241864 x at 2.607759e-04 7.8628284 209946 at 3.394349e-04 7.5841819 226134 s at 3.394349e-04 7.5796142 219049 at 3.394349e-04 7.5611453 212230 at 3.437873e-04 7.5144622 202242 at 3.437873e-04 7.4692929 206343 s at 3.437873e-04 7.4351813 209739 s at 3.437873e-04 7.4102897 218162 at 3.437873e-04 7.3952541 207802 at 3.628879e-04 7.3277217 1557883 a at 3.960407e-04 7.2116284 224901 at 3.970722e-04 7.1634061 201720\_s\_at 3.970722e-04 7.1632751 212543 at 3.970722e-04 7.1329812 227963 at 3.970722e-04 7.1309309 226571 s\_at 3.970722e-04 7.1180314 213436\_at 4.048280e-04 7.0836317 1553995 a at 4.086780e-04 7.0237673 203910 at 4.086780e-04 7.0092773 227156 at 4.163738e-04 6.9768532 228863 at 4.312610e-04 6.9299267 223395 at 4.436215e-04 6.8893262 206068 s at 4.488402e-04 6.8637730 208891 at 4.602533e-04 6.8117761 212942 s at 4.722021e-04 6.7743172 1553602 at 4.723689e-04 6.7598130 205534 at 4.874195e-04 6.7037560 205767 at 4.910509e-04 6.6834605 228742 at 5.091483e-04 6.6373060 223557 s at 5.154060e-04 6.6130070 235274 at 6.057231e-04 6.4401009 216870\_x\_at 6.125176e-04 6.4171372

```
211964 at 6.454816e-04 6.3295445
225240_s_at 6.454816e-04 6.2924306
243309 at 6.454816e-04 6.2852375
213947 s at 6.454816e-04 6.2581704
219874 at 6.454816e-04 6.2553338
243880 at 6.454816e-04 6.2439405
227020 at 6.454816e-04 6.2438213
224856 at 6.539833e-04 6.2022963
207623 at 6.539833e-04 6.1904884
205404 at 6.539833e-04 6.1678105
202712 s at 6.539833e-04 6.1635948
205110 s at 6.539833e-04 6.1623374
226463 at 6.903252e-04 6.1020620
1560225 at 6.972189e-04 6.0826487
217996 at 7.123440e-04 6.0426429
217997 at 7.183258e-04 6.0200746
221731 x at 7.183258e-04 6.0136493
214803 at 7.183258e-04 5.9951743
213201 s at 7.183258e-04 5.9325910
210663 s at 7.183258e-04 5.9311777
226921 at 7.183258e-04 5.9240690
224691 at 7.183258e-04 5.9194505
222920 s at 7.183258e-04 5.9133522
208892 s at 7.204667e-04 5.9018388
227061 at 7.673868e-04 5.8087576
201150 s at 7.879911e-04 5.7672933
203836 s at 8.134517e-04 5.6969586
226757_at 8.294682e-04 5.6707515
205476 at 9.146580e-04 5.5401028
209355 s at 9.496562e-04 5.4777211
203837 at 9.496562e-04 5.4747305
204194_at 9.582670e-04 5.4515579
213316 at 9.832287e-04 5.4035676
226189 at 1.001034e-03 5.3632694
215629 s at 1.001034e-03 5.3606124
210503 at 1.005477e-03 5.3444690
227486 at 1.005477e-03 5.3426781
214467 at 1.018248e-03 5.3224214
235199 at 1.018248e-03 5.2857763
209687 at 1.018248e-03 5.2840786
205003 at 1.031830e-03 5.2650058
227289 at 1.067287e-03 5.2200302
204620_s_at 1.101041e-03 5.1776910
```

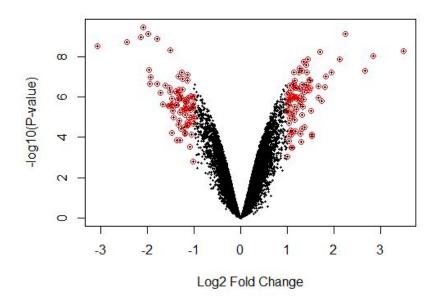
```
226370 at 1.106059e-03 5.1608199
210118 s at 1.153201e-03 5.0688550
204385 at 1.153201e-03 5.0681904
204533 at 1.173002e-03 5.0268347
226302 at 1.217714e-03 4.9682196
213391 at 1.267618e-03 4.9002187
206290 s at 1.267618e-03 4.8819910
204114 at 1.276420e-03 4.8622360
211506 s at 1.288838e-03 4.8403908
228033 at 1.299364e-03 4.8195480
223869 at 1.341506e-03 4.7542758
224938 at 1.341506e-03 4.7433107
238937 at 1.357385e-03 4.7221270
219287 at 1.358013e-03 4.6899355
226218 at 1.462181e-03 4.5839554
239866 at 1.464867e-03 4.5640009
240391 at 1.485814e-03 4.5368289
227875 at 1.507949e-03 4.5136479
212226 s at 1.570247e-03 4.4553394
1557882_at 1.750037e-03 4.2966147
1553994_at 1.779576e-03 4.2360641
222446 s at 1.902143e-03 4.1062131
238953 at 1.916790e-03 4.0876909
227480_at 1.964073e-03 4.0601861
211122 s at 1.986451e-03 4.0267989
204597 x at 2.020393e-03 3.9967093
1552546 a at 2.034249e-03 3.9697950
226492_at 2.106711e-03 3.9109463
208893 s at 2.110821e-03 3.9056727
1553684 at 2.466869e-03 3.7062751
204619 s at 2.548792e-03 3.6445963
238462 at 2.863485e-03 3.4667162
219961 s at 3.051170e-03 3.3805065
231568 at 3.096466e-03 3.3461575
1557094 at 3.234769e-03 3.2692234
210941 at 3.292331e-03 3.2096335
228450 at 3.365718e-03 3.1739324
209839 at 3.586247e-03 3.0775943
207173 x at 3.826645e-03 2.9324569
211959 at 3.998774e-03 2.8595595
222662 at 4.290230e-03 2.7236000
1569157 s at 4.404616e-03 2.6756776
224367_at 4.907541e-03 2.4765306
```

```
205830 at 5.000643e-03 2.4525107
212706_at 5.101094e-03 2.4149637
205535_s_at 5.580465e-03 2.2430129
210355 at 5.586227e-03 2.2378631
205532 s at 5.808908e-03 2.1780336
1558501_at 5.837822e-03 2.1496409
201147_s_at 6.099969e-03 2.0657510
1555673 at 6.197900e-03 2.0430339
235561 at 6.275685e-03 2.0162314
202920 at 6.724926e-03 1.8978896
243009 at 6.913504e-03 1.8528277
220030 at 8.990789e-03 1.3947082
206300_s_at 9.250357e-03 1.3480368
205533_s_at 9.416580e-03 1.3192648
220301 at 1.041079e-02 1.1586786
1554131 at 1.420825e-02 0.6515165
[ reached 'max' / getOption("max.print") -- omitted 5 rows ]
upregulated <- differentially_expressed[differentially_expressed$logFC > 0,]
genes of interest <- AnnotationDbi::select(hgu133plus2.db,
                       keys=rownames(upregulated),
                       columns=c("SYMBOL","ENTREZID","GENENAME"),
                       keytype="PROBEID")
genes_of_interest
#save data into a csv file
write.csv(genes_of_interest,'genes_of_interest.csv')
#volcanoplots
volcanoplot(fitted.ebayes)
```



interesting\_genes <- topTable(fitted.ebayes,number=Inf,p.value = 0.1,lfc=1)
volcanoplot(fitted.ebayes, main=sprintf("%d features pass our
cutoffs",nrow(interesting\_genes)))</pre>

### 171 features pass our cutoffs

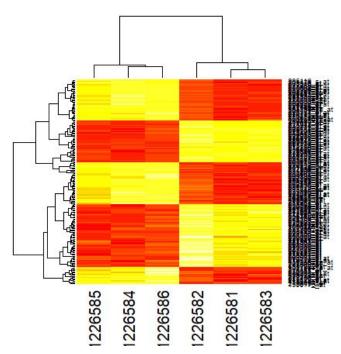


points(interesting\_genes[['logFC']],-log10(interesting\_genes[['P.Value']]),col='red')

#### #heatmaps

#### #normalise expression value

eset\_of\_interest <- eset[rownames(interesting\_genes),]
heatmap(exprs(eset\_of\_interest))</pre>



#fix and beautify heatmap

install.packages("RColorBrewer")

library(RColorBrewer)

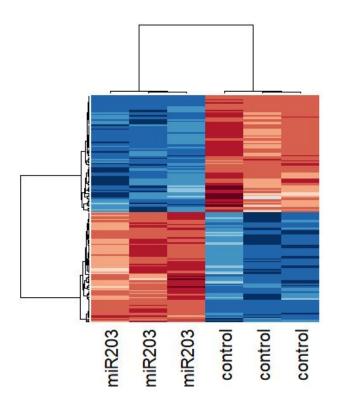
eset\_of\_interest <- eset[rownames(interesting\_genes),]</pre>

heatmap(exprs(eset\_of\_interest),

labCol=eset\$culture,labRow=NA,

col = rev(brewer.pal(10, "RdBu")),

distfun = function(x) as.dist(1-cor(t(x))))



#### #Results

recommendations <- AnnotationDbi::select(hgu133plus2.db,

keys=rownames(interesting\_genes), columns=c("SYMBOL","ENTREZID","GENENAME"), keytype="PROBEID")

write.csv(recommendations,'recommendations.csv')

#shown in appendix below